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3 HUMAN INFLUENCES ON THE CLIMATE

Chapter Summary:

The global climate is naturally variable on all time scales. Anthropogenic CO₂ emissions add to that variability by changing the total radiative energy balance in the atmosphere.

The IPCC has downplayed the role of the sun in climate change but there are plausible solar irradiance reconstructions that imply it contributed to recent warming.

Climate projections are based on IPCC emission scenarios that have tended to exceed observed trends. Most academic climate impact studies in recent years are based upon the extreme RCP 8.5 scenario that is now considered implausible; its use as a business-as-usual scenario has been misleading.

Carbon cycle models connect annual emissions to growth in the atmospheric CO₂ stock. While models disagree over the rate of land and ocean CO₂ uptake, all agree that it has been increasing since 1959.

There is evidence that urbanization biases in the land warming record have not been completely removed from climate data sets.

3.1 Components of radiative forcing and their history

3.1.1 Historical radiative forcing

A changing climate has been the norm throughout the Earth's 4.6-billion-year history. The Earth's temperature and weather patterns change naturally over timescales ranging from decades to millions of years. Natural variations in the surface climate originate in two ways. Internal climate fluctuations associated with circulations in the atmosphere and ocean exchange energy, water, and carbon between the atmosphere, oceans, land, and ice. External influences on the climate system include variations in the energy received from the sun and the effects of volcanic eruptions. Human activities influence climate through changing land use and land cover. Humans are also changing the composition of the atmosphere by emissions of CO₂ and other greenhouse gases and by altering the concentration of aerosol particles in the atmosphere.

The earth is warmed by the sunlight it absorbs and is cooled by the heat it radiates to space. Averaged over the Earth's surface, each of these processes involve power flows of about 240 Watts per square meter (W/m²). When they are in balance, there are no net external causes of warming or cooling. Both human and natural influences on the climate alter this balance and so cause the climate to change.

Influences on the Earth's energy balance at the top of the atmosphere are quantified by "radiative forcing", the extent to which they disturb the warming/cooling balance; a positive forcing warms while a negative forcing cools. The IPCC's estimated history of major components of radiative forcing since 1750 is shown in the following two figures from its AR6.

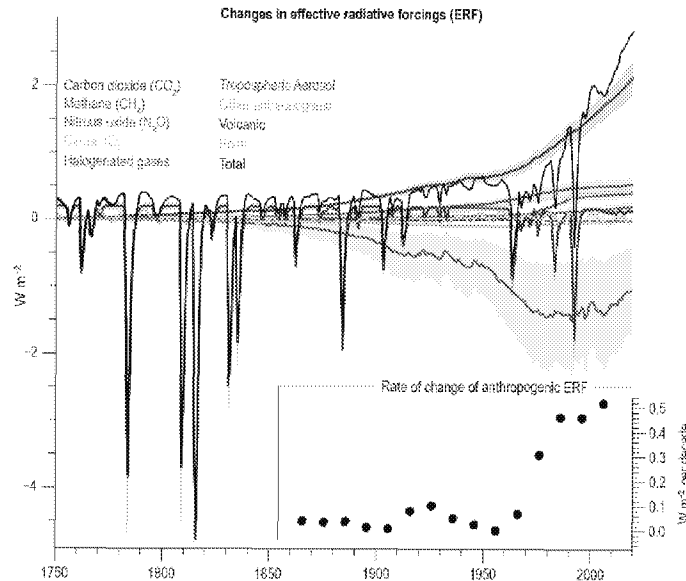


Figure 3.1.1: IPCC estimates of radiative forcing components over time. Shading indicates uncertainty ranges. Source: AR6 WGI Ch2 Fig. 10

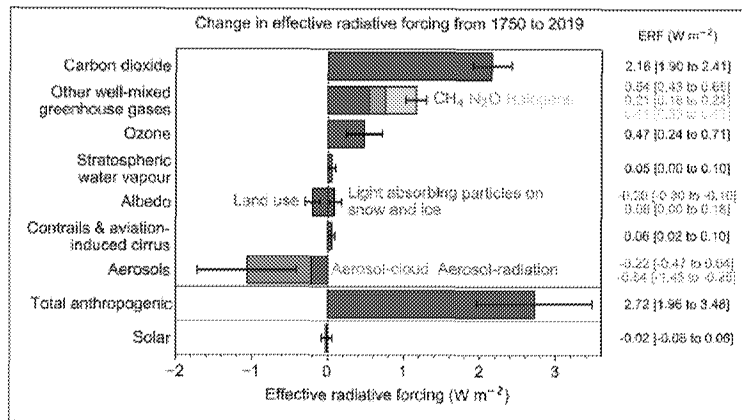


Figure 3.1.2 IPCC estimates of radiative forcing component changes from 1750 to 2019. Source: AR6 WGI Ch 7 Fig. 7-6.

These graphs show that the total radiative forcing is comprised of both natural and anthropogenic components. Carbon dioxide is the largest human influence on the climate and the one most relevant to the influence of fossil fuel use. It exerts a warming influence by decreasing the cooling power of the atmosphere. Emissions of CO₂ are accumulating in the atmosphere, as described in the following section, so that the warming influence is growing. Other greenhouse gases (methane, nitrous oxide, halogens, and ozone) act similarly, currently adding another 75 percent to CO₂'s warming. Aerosols exert an overall cooling effect, although with large uncertainties in the way they catalyze the formation of reflective clouds. As a result, understanding the causes of recent warming requires not just identifying the warming effects of CO₂, but also the more uncertain cooling effects of aerosols.

The IPCC assesses the change in the radiative forcing by the sun to be negligible, based on their preference for data reconstructions that imply minimal solar change since preindustrial times. But Connolly *et al.* (2021) reviewed sixteen different Total Solar Irradiance (TSI) reconstructions in the literature

covering the years 1600-2000; the reconstructions vary from almost no change in TSI to a relatively large upward trend. Those authors note that the variation in TSI reconstructions combined with variations in surface temperature reconstructions allows for inferences consistent with either no or most 20th century warming being attributable to the sun.

A particularly thorny issue is the gap in TSI data between 1989 and 1991 due to a delay in the launch of a monitor following the Space Shuttle Challenger disaster on January 28 1986. This delay prevented a replacement satellite from being launched in time to overlap with, and its readings to be intercalibrated with, the prior system (Zacharias 2014, Scafetta *et al.* 2019). This is called the ACRIM (Active Cavity Radiometer Irradiance Monitor) gap problem. The question of whether there is an upward trend in TSI over 1978 to 2018 hangs on how the ACRIM data gap is filled. Connolly *et al.* (2021) found that the IPCC's consensus statements on solar forcing were formulated prematurely through the suppression of dissenting scientific opinions.

Another natural radiative forcing component is volcanic aerosols, which exert an episodic cooling influence. Box 4.1 in the IPCC AR6 Report addresses the climate impact of volcanic eruptions, noting three explosive volcanic eruptions that occurred in the first half of the 19th century. This included the 1815 Tambora eruption that resulted in the 'year without summer', with multiple harvest failures across the Northern Hemisphere. There is uncertainty about the sign of the relatively small forcing due to the submarine volcano Hunga Tonga which erupted in 2022 (Jenkins *et al.* 2023, Schoeberl *et al.* 2024).

Figure 3.1.1 shows that the anthropogenic forcing component was negligible before about 1900 and has increased steadily since, rising to almost 3 W/m² today. However, this is still only about 1 percent of the unperturbed radiation flows, making it a challenge to isolate the effects of anthropogenic forcing; state-of-the-art satellite estimates of global radiative energy flows are only accurate to a few W/m².

Natural sources of global energy imbalance other than volcanoes and total solar irradiance (TSI) are not included in these graphs because they remain largely unknown.

3.1.2 Change in atmospheric CO₂ since 1958

Carbon dioxide's warming influence depends on how much "extra" CO₂ accumulates in the atmosphere- *i.e.*, its concentration above the preindustrial value of 280 ppm. The CO₂ level as recorded at the Mauna Loa observatory in Hawaii, generally used as the representative global average concentration, is available online at [[HYPERLINK "https://gml.noaa.gov/ccgg/trends/index.html"](https://gml.noaa.gov/ccgg/trends/index.html)]. The concentration was about 316 ppm at the start of the record in 1959 and is now about 430 ppm, a 36 percent increase. At the end of the last glaciation CO₂ levels had fallen to about 180 ppm. As discussed in Chapter 2, C3 plants begin dying at CO₂ levels below about 140 ppm and C4 plants at levels below 100 ppm, so if CO₂ levels had continued falling plant life would have been imperiled.

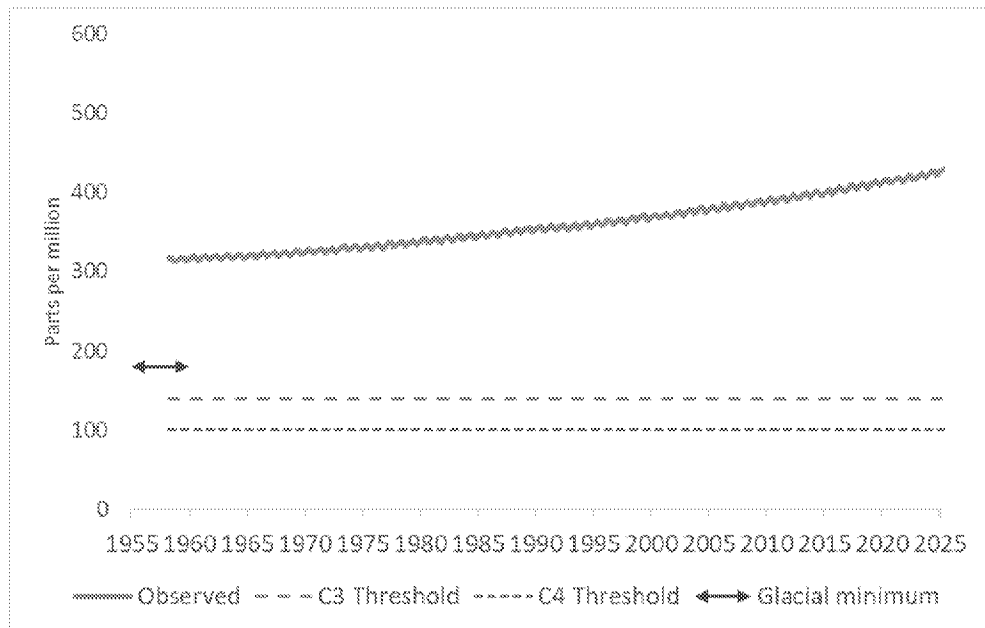


Fig. 3.1.3. Yearly average atmospheric CO₂ concentrations (1959-2025) in ppm measured at Mauna Loa (blue). C3 Threshold: Level below which C3 plants begin dying (140 ppm, see Chapter 2). C4 Threshold: Level below which C4 plants begin dying (100 ppm, see Chapter 2). Glacial minimum: Minimum level during recent glaciations (purple arrow). CO₂ data source: [HYPERLINK "<https://gml.noaa.gov/ccgg/trends/index.html>"]

The annual increase in concentration is only about half of the CO₂ emitted because land and ocean processes currently absorb “excess” CO₂ at a rate approximately 50 percent of the human emissions. Future concentrations, and hence future human influences on the climate, therefore depend upon two components: (1) future rates of global human CO₂ emissions, and (2) how fast the land and ocean remove extra CO₂ from the atmosphere. We discuss each of these in turn.

3.2 Future emission scenarios and the carbon cycle

3.2.1 Emission scenarios

Assessing the dangers of future GHG emissions requires assumptions about what those emissions will be. Future emissions, and hence human influences on the climate, will depend upon future demographics, economic activity, regulation, and energy and agricultural technologies. Various assumptions about each of those lead to projections of greenhouse gas emissions and concentrations, aerosol concentrations, and changes in land use, which ultimately can be combined into assumptions about anthropogenic radiative forcing.

The great uncertainties about these many factors make it impossible to precisely predict future emissions. Instead, the IPCC has used various sets of scenarios meant to span a plausible range of possibilities for population, economy, and technologies. Recent versions of the scenarios are labeled by a number indicating the anthropogenic radiative forcing expected in 2100 under that scenario. Thus, a scenario labeled with a “6” corresponds to 6 W/m² of human-induced radiative forcing (warming) at the end of the century. (Recall current anthropogenic radiative forcing is about 2.7 W/m².)

Although the IPCC does not claim its emission scenarios are forecasts, they are often treated as such. Comparisons of past scenario groups against observations show that IPCC emission projections have tended to overstate actual subsequent emissions. For the IPCC Third and Fourth Assessment Reports a set of emission projections from the *Special Report on Emission Scenarios* was used; these were referred to as the SRES scenarios. McKittrick *et al.* (2012) showed that, when converted to per capita values, the SRES scenario emissions distribution was skewed upwards compared with observed trends. The bias of the SRES scenarios was confirmed by the later analysis of Hausfather *et al.* (2019) who showed that observed atmospheric CO₂ concentrations tracked the low end of the SRES range and also of subsequent IPCC scenario ranges (Figure 3.2.1).

For AR5 the IPCC developed a new set of scenarios called the Representative Concentration Pathways (RCPs). These were identified by a number representing the increase in forcing and were thus called RCP2.6, RCP4.5, RCP6.0 and RCP8.5. RCP2.6 (implying an anthropogenic radiative forcing in 2100 of 2.6 W/m²) describes a GHG concentration pathway leading to warming well below 2°C. At the other end of the scale RCP8.5 is an extreme outcome implying nearly 5°C warming from 1900 to 2100.

RCP8.5 came to be referred to as the no-policy baseline, or “business-as-usual” scenario in both the academic literature and popular media. It was therefore used to generate the reference outcome supposedly representing the 21st century world in the absence of increasingly stringent emission reduction policies. But RCP8.5 was intended as a low-probability high emissions scenario and its use as a business-as-usual baseline has been criticized as grossly misleading.¹ Hausfather and Peters (2020a) writing in a commentary in *Nature*, pointed out that RCP8.5 was developed as an extreme worst-case, and its misuse as a “business as usual” baseline has resulted in a large number of misleading studies and media reporting.

The implausibility of the RCP8.5 scenario was examined by Burgess *et al.* (2021). The implausibility of RCP8.5 should not be interpreted as *very unlikely* (e.g. 95th percentile) or a “worst case”, but rather as genuinely implausible owing to the implausibility of the inputs required to reach a forcing of 8.5 W/m². They noted that RCP8.5 has already diverged from observed trends in energy use and the near future trends diverge sharply from those of the International Energy Agency (IEA), which provides market-based projections of energy use for the coming decades. Pielke Jr. *et al.* (2022) further showed that the historic and projected IEA trends run near the bottom of the envelopes of both RCP projections and the more recent Shared Socioeconomic Pathway (SSP) scenario trends.

Schwalm *et al.* (2020) defended the use of RCP8.5 on the grounds that cumulative CO₂ emissions over 2005-2020 track it more closely than the lower RCP scenarios. They also argue that a modified version of the IEA scenarios closely track RCP8.5 in the coming decades. Hausfather and Peters (2020b) responded that the skill of RCP8.5 over those 15 years is due to offsetting errors in its representation of CO₂ from fuel use and land use change, and the apparent agreement with IEA in coming decades is due to Schwalm *et al.* adding in very high land use emissions. The IEA’s own projected CO₂ emissions track well below RCP8.5.

Widespread use of RCP8.5 as a no-policy baseline has created a bias towards alarm in the climate impacts literature. The extent of this problem was confirmed in a literature analysis by Pielke Jr. and Ritchie (2020). They found that some 16,800 scientific papers published between 2010 and 2020 used the RCP8.5 scenario, with about 4,500 of the articles linking RCP8.5 to the concept of “business-as-usual”. Their analysis showed how RCP8.5 was misused not only by individual researchers, but also by influential science agencies like the IPCC and the U.S. National Climate Assessment (USNCA), which has directly led to misleading coverage in prominent media outlets.

¹ This extreme scenario is useful for modelers, since a large forcing generates a large response (warming) making it easier to assess a model’s sensitivity. But that’s very different than claiming it is a plausible future outcome.

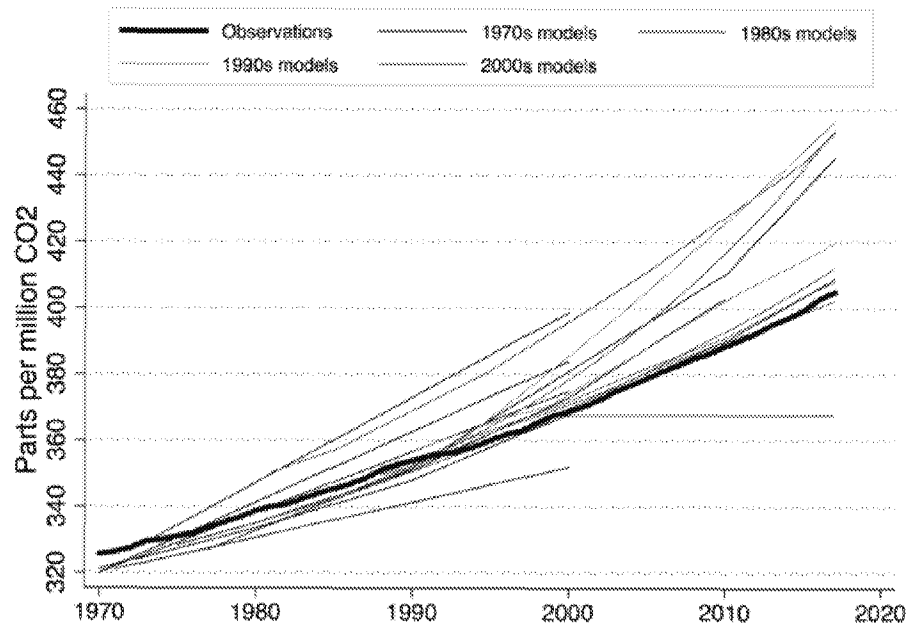


Figure 3.2.1 Since the 1970s successive families of emission and concentration projections (colored lines) have consistently overestimated observations (black line). Source: Hausfather *et al.* (2019) Figure S4.

Pielke and Ritchie (2020) reported new studies using RCP8.5 were published at a rate of about 20 per day, with about two per day specifically linking RCP8.5 and “business as usual.” They conclude that the climate research community has spent a decade “committing scientific resources to science fiction” and that “The scientific literature has become imbalanced in an apocalyptic direction.”

The IPCC developed a new set of scenarios for AR6, the “Shared Socioeconomic Pathway” (SSP) scenarios, which have continued the bias shown in the RCP and SRES scenarios. Figure 3.2.2 shows total global observed CO₂ emissions compiled by the International Energy Agency (IEA) merged with the emission projection from the IEA taking account of energy use projections and current policies. The other lines show the range of IPCC SSP scenarios (SSP1-SSP5). As of 2023, global CO₂ emissions have been tracking well below SSP7.0 and are even below SSP2-4.5.

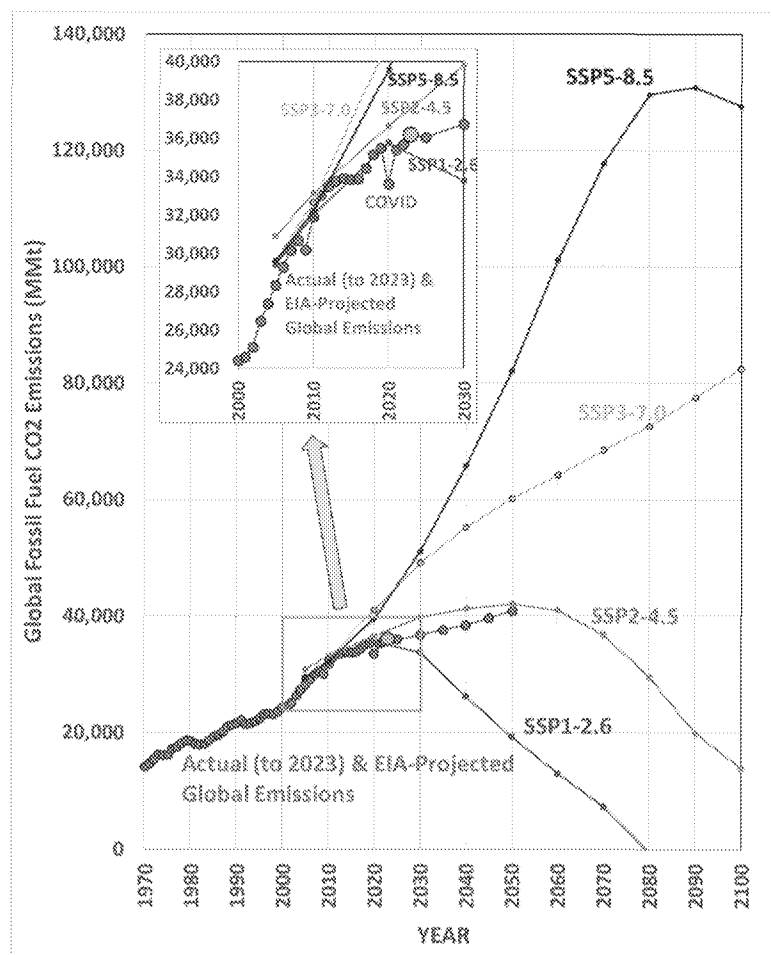


Figure 3.2.2. Observed and projected CO₂ emissions. Source: IPCC (SSP scenarios) and Energy Information Administration (EIA). Green: observed historical emissions and EIA projections. Other lines: SSP1-5. Data source: Friedlingstein *et al.* (2024).

3.2.2 The carbon cycle relating emissions and concentrations

Carbon dioxide emissions from fossil fuel burning (and to a lesser extent deforestation and cement production) have led to steadily increasing CO₂ concentrations in the atmosphere, as shown in Fig. 3.1.3. The relation between emissions and concentration is determined by the global carbon cycle of land and ocean processes that exchange carbon with the atmosphere. Our understanding of these processes was reviewed by Crisp *et al.* (2021).

There are about 850 Gt of carbon (GtC) in the Earth's atmosphere², almost all of it in the form of CO₂. Each year, biological processes (plant growth and decay) and physical processes (ocean absorption and outgassing) exchange about 200 GtC of that carbon with the Earth's surface (roughly 80 GtC with the land and 120 GtC with the oceans). Before human activities became significant, removals from the atmosphere were roughly in balance with additions. But burning fossil fuels (coal, oil, and gas) removes carbon from the ground and adds it to the annual exchange with the atmosphere. That addition (together with a much

² Because CO₂ is chemically transformed through the course of the carbon cycle, it is more convenient to track atoms of carbon rather than molecules of CO₂. One gigatonne of carbon (GtC) is equivalent to about 3.7 Gt of CO₂.

smaller contribution from cement manufacturing) amounted to 10.3 GtC in 2023, or only about 5 percent of the annual exchange with the atmosphere.

The carbon cycle accommodates about 50 percent of humanity’s small annual injection of carbon into the air by naturally sequestering it through plant growth and oceanic uptake, while the remainder accumulates in the atmosphere (Ciais *et al.*, 2013). For that reason, the annual increase in atmospheric CO₂ concentration averages only about half of that naively expected from human emissions.

To project future CO₂ concentrations in the atmosphere, and hence future human influences on the climate, it is important to know how the carbon cycle might change in the future. The historical near constancy of that 50 percent fraction means that the more CO₂ humanity has produced, the faster nature removed it from the atmosphere. That 50 percent fraction changes somewhat from year to year due to natural carbon cycle imbalances from El Nino, La Nina, and varying weather patterns. There was also a substantial additional reduction in atmospheric CO₂ after the 1991 eruption of Mt. Pinatubo, a curious result that has yet to be explained (Angert *et al.*, 2004).

The main processes that remove excess CO₂ from the atmosphere are increased growth of land vegetation (especially at high latitudes), some increase in the sequestering of carbon in soils, and uptake of CO₂ by the ocean due to the increasing partial pressure of atmospheric CO₂ over that of CO₂ dissolved in the ocean. All twenty land carbon cycle models tracked by the Global Carbon Project (Friedlingstein *et al.*, 2024) show land processes have been removing excess CO₂ at an increasing rate since 1959. This is consistent with a “global greening” phenomenon (Chapter 2.1) observed by satellites since monitoring of global greenness began in 1982.

While land vegetation has been responding positively to more atmospheric CO₂, uptake of extra CO₂ by ocean biological processes remains too uncertain to be measured reliably. Our current understanding of these and many more carbon cycle processes was reviewed by Crisp *et al.* (2021).

CO₂ uptake by land processes

The uptake of extra CO₂ from the atmosphere by land surface processes (as also inferred from global greening) has been modeled with 20 different dynamic global vegetation models, the outputs of which are updated every year by the Global Carbon Project (Friedlingstein, 2024). As seen in Fig. 3.2.3, all of those models agree that vegetation and soils have been sequestering carbon from the atmosphere. But we also see that the long-term trends over 1959 to 2023 (65 years) vary widely between models, by nearly a factor of 7. This demonstrates that there remains considerable uncertainty in how fast land processes are removing CO₂ from the atmosphere, which in turn creates uncertainty in future atmospheric CO₂ concentrations, which then produce uncertainty in climate model simulations of future climate change.

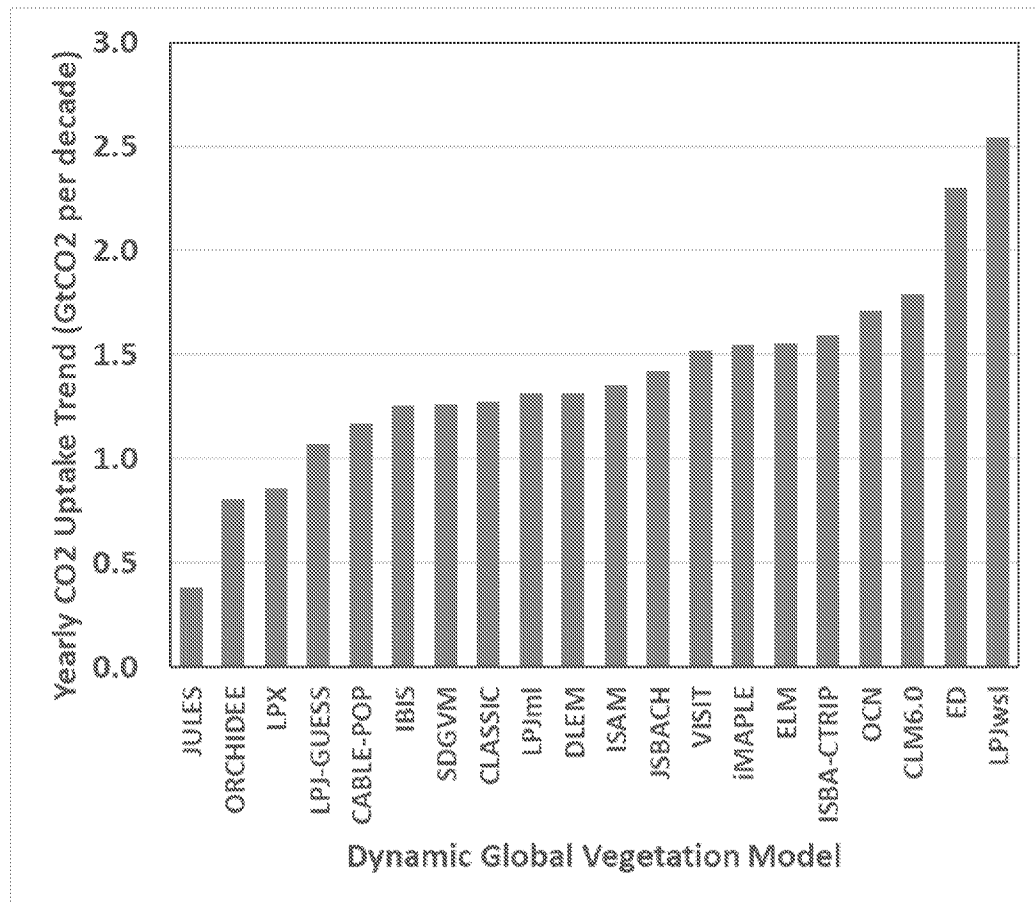


Figure 3.2.3 Trends of annual CO₂ uptake (GtCO₂ per year per decade) by land processes during 1959-2023 simulated by 20 different dynamic global vegetation models periodically reported by the Global Carbon Project (Friedlingstein, 2024).

CO₂ uptake by ocean processes

The uptake of extra CO₂ from the atmosphere by ocean processes has been modeled with 10 different ocean biogeochemistry models, the outputs of which are updated every year by the Global Carbon Project (Friedlingstein, 2024). Like the results from the land models, all of the ocean models agree that the global oceans have been sequestering carbon from the atmosphere at an increasing rate during 1959-2023 (Fig. 3.2.4). Unlike the land models, however, the ocean models show much better agreement with each other, with the model producing the fastest increasing CO₂ uptake being only 65 percent faster than the model with the most slowly increasing CO₂ uptake. In spite of the relative agreement among models, Friedlingstein *et al.* (2022) notes that there is substantial discrepancy between the different methods on the strength of the ocean sink over the last decade, particularly in the Southern Ocean.

Note that the average trend in CO₂ uptake across all land models in Fig. 3.2.3 is 25 percent larger than the average trend in ocean uptake. This suggests land processes are increasing in their ability to remove CO₂ faster than ocean processes are increasing their CO₂ sequestration.

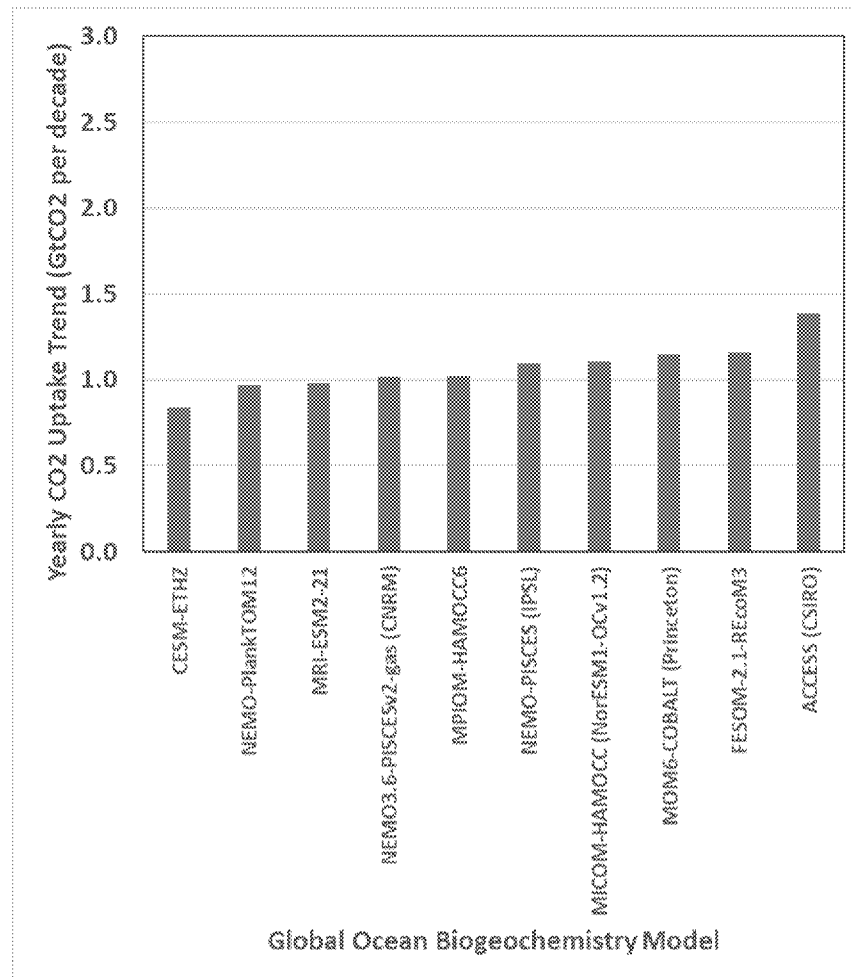


Figure 3.2.4 Trends of annual CO₂ uptake (GtCO₂ per year per decade) by ocean processes during 1959-2023 simulated by 10 different ocean biogeochemistry models periodically reported by the Global Carbon Project (Friedlingstein, 2024).

3.3 Urbanization influence on temperature trends

Historical temperature data over land has been collected mainly where people live. This raises the problem of how to filter out non-climatic warming signals due to Urban Heat Islands (UHI) and other changes to the land surface. If these are not removed the data might over-attribute observed warming to greenhouse gases. The IPCC acknowledges that raw temperature data are contaminated with UHI effects but claims to have data cleaning procedures that remove them. It is an open question whether those procedures are sufficient.

AR6 downplayed this issue by saying (WGI p. 235) that no recent evidence had emerged to alter the AR5 finding that urbanization causes an upward bias of no more than 10 percent in the global land surface warming trend. AR5 (WGI p. 189) also cited the 10 percent upper bound without citing a source. AR4 (WGI p. 244) cited Jones *et al.* (1990) and Peterson *et al.* (1999) as the basis of the claim. Peterson *et al.* had failed to find any difference in trends between rural and urban samples, although their definition of rural included local populations up to 10,000 persons while the relative influence of urbanization begins well below that (Spencer *et al.*, 2025). Jones *et al.* compared rural/urban warming in three regions: Eastern Australia, Eastern China and Western Soviet Union. Their definition of “rural” included towns of up to

10,000 in the former Soviet Union and up to 100,000 in China. They found relative warming biases greater than 10 percent in these areas but conjectured that the urbanization effect averaged over the areas they did not examine would bring the global land bias to under 10 percent of the observed warming trend.

Several papers appeared in print prior to the IPCC AR4 that argued that the warming effect of UHIs added a relatively large (30-50%) component to observed warming and was not simulated by climate models (de Laat and Maurellis 2006, McKittrick and Michaels 2007). These findings were based on correlations between locations of maximum warming over land with locations of maximum socioeconomic development. AR4 asserted (p. 244) that these correlations were an artefact of natural atmospheric circulations and were in fact statistically insignificant, and on that basis set the findings aside. Their claim was controversial because it was presented with no supporting evidence. McKittrick (2010) and McKittrick and Nierenberg (2010) showed that taking into account various conjectured alternative explanations for the correlations did not affect their significance. AR5 (p. 189) conceded that AR4 had provided “no explicit evidence” for its assessment and further acknowledged, based on these papers, that there was “significant evidence for such contamination of the record” i.e. a warming bias in the land record. However as already noted, elsewhere in the AR5 report they carried forward AR4 claim that it was less than 10 percent of observed warming. Further they provided no caution about using the land record for climate measurement despite conceding the evidence for UHI contamination. Recently Soon *et al.* (2023) estimated an urbanization bias in the Northern Hemisphere land record over 1850-2018 sufficient to increase the trend in the blended record from 0.55°C to 0.89°C per century.

Some studies providing evidence against UHI contamination compared warming rates between rural and urban locations (Jones *et al.* 1990, Peterson *et al.* 1999, Wickham *et al.* 2013). It is not known whether such methods would be capable of detecting UHI bias even when present. The influence of UHI warming is logarithmic in population, in other words it is strongest at low population density then levels off as local urbanization expands (Oke 1973, Spencer *et al.* 2025). Hence failure to find a difference in warming rates between urban and rural stations does not prove the absence of UHI contamination. McKittrick (2013) provided an empirical demonstration in which the rural/urban trends were not significantly different in a data set shown on other grounds to be contaminated with UHI bias.

Parker (2006) examined a sample of urban locations and found no difference in trends between subsets partitioned according to nighttime wind speed, concluding on this basis that urbanization could not be a significant factor. Here again the question is whether such a method would find UHI bias even if present. McKittrick (2013) presented an example in which UHI-contaminated data did not exhibit significant trend differences when stratified according to wind speed.

The challenge in measuring UHI bias is relating local temperature change to a corresponding change in population or urbanization, rather than to a static classification variable such as rural or urban. Spencer *et al.* (2025) used newly available historical population archives to undertake such an analysis and found evidence of significant UHI bias in U.S. summertime temperature data.

In summary, while there is clearly warming in the land record, there is also evidence that it is biased upward by patterns of urbanization and that these biases have not been completely removed by the data processing algorithms used to produce climate data sets.

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PART II: CLIMATE RESPONSE TO CO₂ EMISSIONS

4 CLIMATE SENSITIVITY TO CO₂ FORCING

Chapter Summary

There is growing recognition that climate models are not fit for the purpose of determining the Equilibrium Climate Sensitivity (ECS) of the climate to increasing CO₂. The IPCC has turned to data-driven approaches including historical data and paleoclimate reconstructions, but their reliability is diminished by data inadequacies.

Data-driven ECS estimates tend to be lower than climate model-generated values. The IPCC AR6 upper bound for the likely range of ECS is 4.0°C, lower than the AR5 value of 4.5°C. This lowering of the upper bound seems well justified by paleoclimatic data. The AR6 lower bound for the likely range of ECS is 2.5°C, substantially higher than the AR5 value of 1.5°C. This raising of the lower bound is less justified; evidence since AR6 finds the lower bound of the *likely* range to be around 1.8°C.

4.1 Introduction

The magnitude of the climate's response to increasing concentrations of CO₂ is central to the scientific debate on anthropogenic climate change, and so also to the public debate on “climate action.” The simplest measure of that response is the rise in the global average surface temperature, quantified by the Equilibrium Climate Sensitivity (ECS). ECS is defined as the amount of warming expected in response to a doubling of CO₂ from its pre-industrial concentration of 280 ppm, after all climate components have had time to adjust. Some components, like temperatures in the lower atmosphere (troposphere), adjust rapidly, while others such as the deep ocean and cryosphere might take as long as centuries. A related measure, the Transient Climate Response (TCR), better describes the shorter time scales; it is defined as the amount of warming when the CO₂ concentration is doubled by rising one percent annually for 70 years.

The 1979 Charney Report for the U.S. National Academy of Sciences (National Research Council 1979) proposed that the most likely ECS was $3.0 \pm 1.5^\circ\text{C}$. The IPCC repeatedly reaffirmed that range with only minor variations until its most recent AR6. AR5 termed 1.5–4.5°C as the *likely* range (66 percent probability) and stated that ECS is *extremely unlikely* (95 percent probability) to be below 1.0°C and *very unlikely* (90 percent probability) to exceed 6.5°C.

The uncertainty in ECS has remained stubbornly wide, despite many individual studies that claimed to narrow it (Hausfather 2023). Most recently, AR6 narrowed the *likely* range to 2.5–4.0°C and deemed the *very likely* range to be 2.0–5.0°C. This narrowing on the low end is disputed, as will be discussed below.

Uncertainties in ECS are highly consequential for policy making. As will be discussed in Chapter 11, economic models use ECS values to project the costs of CO₂ emissions. The traditional value (3.0 °C) has typically yielded modest global social costs of CO₂ emissions, sufficient to justify some policy actions, but mostly deferred to later in this century. If ECS is very high (above 4.5°C) immediate aggressive emission controls become more imperative, whereas no CO₂ emission controls are economically justifiable for ECS below 2.0°C (Dayaratna *et al.* 2017, 2020). Obtaining a precise estimate is impossible, so policy making needs to account for the uncertainty.

By itself, the equilibrium warming effect of a doubling of atmospheric CO₂ is slightly more than 1°C (Soden and Held 2006). Larger values of ECS arise from positive feedbacks that amplify the CO₂ warming. Water vapor feedback is positive: a warmer atmosphere might have more water vapor, which itself is a powerful greenhouse gas. Warmer temperatures also result in less snow and sea ice cover, allowing the Earth to absorb more of the sun's radiation. Some simple estimates of these feedbacks increase the ECS to around 2°C (Sherwood *et al.*, 2020). Larger values of ECS are associated with positive cloud feedbacks.

Climate scientists use multiple lines of evidence to determine the Equilibrium Climate Sensitivity:

- Climate model simulations
- Historical observations
- Paleoclimatic reconstructions
- Process understanding of feedbacks

4.2 Model-based estimates of climate sensitivity

The ECS ranges given in IPCC AR4 and AR5 were obtained primarily by examining the behavior of large-scale climate models, also called General Circulation Models (GCMs). However, the IPCC changed course in its AR6 when it turned to a more direct data-driven methodology. Here we discuss some of the pitfalls of using GCMs to try to determine the Earth's climate sensitivity.

ECS can be determined from climate model simulations by doubling the concentration of CO₂ and allowing several centuries for the warming to equilibrate. To avoid the need for such long simulations, “effective climate sensitivity” is commonly evaluated from a 150-year simulation in response to a sudden quadrupling of CO₂.

In principle, ECS is an emergent property of GCMs—that is, it is not directly parameterized or tuned but rather emerges in the results of the simulation. Otherwise plausible GCMs and parameter selections have been discarded because of perceived conflict with an expected warming rate, or aversion to a model's climate sensitivity being outside an accepted range (Mauritsen et al. 2012). This practice was commonplace for the models used in AR4; modelers have moved away from this practice with time. However, even in a CMIP6 model, Mauritsen and Roeckner (2020) state the following regarding their Max Planck Institute (MPI) climate model (emphasis added):

“We have documented how we tuned the MPI-ESM1.2 global climate model to match the instrumental record of warming; an endeavor which has clearly been successful. Due to the historical order of events, the choice was to do this practically by **targeting an ECS of about 3 K using cloud feedbacks**, as opposed to tuning the aerosol forcing.”

In other words, the MPI modelers chose an ECS value of 3°C and then tuned the cloud parameterizations to match their intended result.

As noted, direct warming from CO₂ doubling is only about 1°C (Soden and Held 2006); further warming arises from climate feedbacks that are not explicitly resolved by the GCM but rely on parameterizations of physical processes. Higher values of ECS arise primarily from positive cloud feedbacks, whereas the magnitude and even the sign of the feedbacks are very uncertain. Elements of cloud feedback include changes in the latitudinal distribution of clouds, changes in the distribution of cloud height (changes in low versus high clouds), changes to the phase of clouds (ice versus liquid), changes in cloud particle size (associated with changes in concentration and/or composition of aerosol particles), changes in the precipitation efficiency of clouds, and even changes in how clouds are distributed over the daily solar cycle (Curry and Webster, 1999). It is difficult for GCMs to simulate any of these processes correctly owing to their small scale, let alone predict how they will change in the future. Further, cloud processes modulate the magnitudes of the water vapor, lapse rate, and the surface albedo feedbacks.

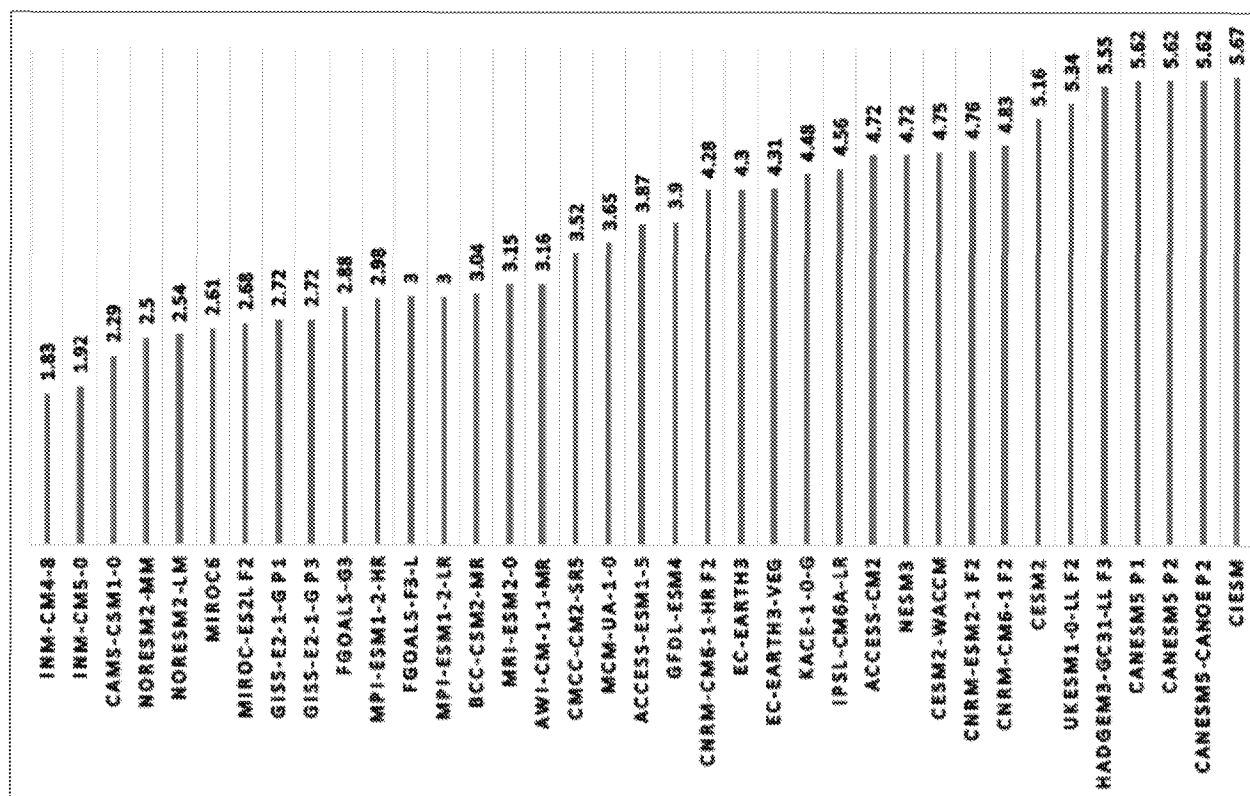


Figure 4.1 Equilibrium Climate Sensitivities in °C of 37 climate models from the CMIP6 ensemble. Identifiers for the various models appear along the horizontal axis. From (Scafetta, 2021)

The spread of ECS values from the CMIP5 ensemble of climate models used in AR5 was 2.0–4.7°C; that range increased for the CMIP6 models used in AR6 to between 1.8 and 5.7°C (Chen *et al.*, 2021, Scafetta 2021, see Figure 4.1). Far from resolving the model-based climate sensitivity the range appears to be growing. The main cause of the overall upward shift in ECS in CMIP6 relative to CMIP5 is a larger positive cloud feedback, driven by changes to the cloud parameterizations in many CMIP6 models (Zelinka *et al.*, 2020)

Because of concerns about model tuning and the high sensitivity to cloud parameterizations, AR6 (2021) did not rely on climate model simulations in their assessment of climate sensitivity, relying instead on data-driven methods.

4.3 Data-driven estimates of climate sensitivity

Climate sensitivity can also be estimated from instrumental records of surface temperatures and ocean heat content, combined with estimates of how climate forcings (e.g., greenhouse gases, solar, volcanoes, aerosols) have changed in the past (Otto *et al.*, 2013). Using this information, a simple empirical Energy Balance Model can be employed. It requires estimating a feedback parameter whose uncertainties are highly amplified in the resulting ECS (Roe and Baker, 2007).

The accuracy of the data-driven methods depends on the quality of the input data. Assumptions are needed about ocean heat storage, and good data is only available for recent decades. The greatest source of uncertainty is the amount and composition of aerosol particles and their interactions with cloud radiative

properties (the so-called aerosol indirect effect; see Figures 3.1.1, 3.1.2). Climate models exhibit warming in response to GHGs but cooling in response to aerosols (Schwartz *et al.*, 2007). Observed 20th century warming can be shown to be consistent either with low ECS and low aerosol cooling, or high ECS and high aerosol cooling. Since fossil fuel use adds both GHGs and aerosols to the atmosphere, both effects need to be estimated to isolate the warming effect of CO₂.

Paleoclimate proxies are also used to evaluate the sensitivity of past climates by comparing paleoclimate changes in the Earth's temperatures to estimates of changes in forcings. The two most informative periods are the last glacial maximum (around 20,000 years ago), which was about 3–7°C colder than today, and a mid-Pliocene period (roughly three million years ago), which was 1–3°C warmer than today. The limits on cooling during the last glacial maximum give the best single evidence that high values of climate sensitivity are unlikely. However, paleoclimate estimates are associated with very large uncertainties in the estimated temperatures and forcings. Further, estimates of climate sensitivity based on past climate states might not be applicable to the current state of the climate system.

A recurring theme in the climate literature is that ECS estimates based on historical data are smaller than ECS estimates inferred from climate models (Sherwood and Forest 2024). About 15 estimates based on historical data appeared in the peer-reviewed literature between 2012 and 2024 yielding ECS best estimates between 1.0°C and 2.5°C, although critics have questioned some of the methods and the data quality. For AR6, the IPCC placed primary weight on the results of Sherwood *et al.* (2020) that combined historical data and paleoclimate proxies with the process-based approach and yielded a best estimate of 3.1°C with a *likely* range of 2.6–3.9°C. Lewis (2022) raised a number of concerns about this result, including methodological errors, outdated input values, and use of subjective Bayesian priors in the analysis. Lewis' analysis found that climate sensitivity is estimated to be much lower and better constrained than in the Sherwood *et al.* analysis – median 2.2°C (1.8–2.7°C in the 17–83 percent *likely* range, and 1.6–3.2°C in the 5–95 percent *very likely* range). The IPCC AR6 estimated only a 5 percent probability that ECS was below 2.3°C, whereas Lewis estimated it to be over 50 percent. The most recent publications on the debate between Sherwood *et al.* and Lewis further defend their respective positions: Sherwood and Foster (2024) and Lewis (2025).

An argument emphasized in AR6 is that data-driven ECS estimates might understate the future warming response to GHGs because of a so-called “pattern effect” (Forster *et al.*, 2021). The tropical Pacific is believed to strongly influence the overall efficiency with which the Earth radiates heat to space, but some regions remove heat more efficiently than others. If the west-to-east temperature gradient in the tropical Pacific is weakened in a warming climate, warming would concentrate where heat is less efficiently removed, raising ECS.

Most climate models simulate that rising GHGs will weaken the west-east temperature gradient, which led the IPCC in AR6 to conclude that data-driven ECS estimates understated the *future* ECS value. However, Seager *et al.* (2019) pointed out that, contrary to models, the west-east temperature gradient has been strengthening over time. They further argued that the mechanism predicting otherwise in climate models was based on a faulty characterization of oceanic dynamics and there is no reason to expect the gradient to weaken. A similar argument was recently made by Lee *et al.* (2024), who concluded that “the trajectory of the observed trend reflects the response to increasing GHG loading in the atmosphere”; in other words, GHG warming should lead to a future strengthening rather than a weakening of the temperature gradient. Increased efficiency of atmospheric cooling implies, if anything, that the future ECS in a warming climate might be *lower* than current estimates.

4.4 Transient Climate Response

The Transient Climate Response (TCR) provides a more useful observational constraint on climate sensitivity. TCR is the global temperature increase that results when CO₂ is increased at an annual rate of 1 percent over a period of 70 years (*i.e.*, doubled gradually). Relative to the ECS, observationally determined values of TCR avoid the problems of uncertainties in ocean heat uptake and the fuzzy boundary in defining equilibrium arising from a range of timescales for the longer-term feedback processes (*e.g.*, ice sheets). TCR is better constrained by historical warming, than ECS. AR6 judged the *very likely* range of TCR to be 1.2–2.4°C. In contrast to ECS, the upper bound of TCR is more tightly constrained. For comparison, the TCR values determined by Lewis (2023) are 1.25 to 2.0°C, showing much better agreement with AR6 values than was seen in a comparison of the ECS values.

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5 DISCREPANCIES BETWEEN MODELS AND INSTRUMENTAL OBSERVATIONS

Chapter summary

Climate models show warming biases in many aspects of their reproduction of the past several decades. In response to estimated changes in forcing they produce too much warming at the surface (except in the models with lowest ECS), too much warming in the lower-and mid-troposphere and too much amplification of warming aloft.

Climate models also produce too much recent stratospheric cooling, invalid hemispheric albedos, too much snow loss, and too much warming in the Corn Belt. The IPCC has acknowledged some of these issues but not all.

5.1 Introduction

Climate models are the primary tool used to project future climate changes in response to increasing atmospheric levels of anthropogenic greenhouse gases. To assess the fitness of climate models for this purpose, it is reasonable to ask how well they reproduce the current climate and its variations over the past century. The box “Climate modeling” gives some detail on how climate models work.

Of great concern is the fact that, after several decades of the climate modeling enterprise involving approximately three dozen models operated by research centers around the world, the range of future warming they produce in response to a hypothetical doubling of atmospheric CO₂ extends over a factor of three, as we discussed in the previous chapter. This range of disagreement among models has not decreased for decades.

Problems with climate models are not just in their disagreement over the future, but also in their ability to replicate the recent past. Here we review some of the most important metrics of climate model accuracy: ability to reproduce historical surface, tropospheric and stratospheric temperature trends; ability to reproduce the vertical warming profile; and ability to reproduce other climate features such as snowfall. In all cases a persistent finding is that models on average err on the side of too much warming in response to estimated historical forcings.

BOX: Climate modeling

All but the simplest climate models represent Earth's land surface using a grid of squares some 100 km wide. To simulate the atmosphere, typically 30 or more grid boxes are stacked above these squares. The ocean is modeled using a similar but finer grid, resulting in tens of millions of grid boxes for the atmosphere and oceans.

The computer models, based on physical laws, calculate how air, water, and energy move between grid boxes over time. The time step can be as small as 10 minutes, and repeating this process millions of times allows the simulation of climate over centuries. Running these models, even on the most powerful supercomputers, can take months. Comparing simulation results with historical climate data helps assess the accuracy of a model, while projections into the future estimate climate changes under assumed human and natural influences.

Despite sounding straightforward, climate modeling is highly complex. Many critical processes occur at scales smaller than the grid size. For instance, the flows of sunlight and heat in the atmosphere depend strongly on cloud cover. Since tracking individual clouds is impractical, researchers must make “subgrid” assumptions about the distribution of clouds in each grid box. Snow and ice cover, which affects how much sunlight is reflected or absorbed by the surface, is another subgrid factor.

Each subgrid assumption requires numerical parameters, which must be carefully set. Modelers initially estimate these parameters based on physics and observed climate patterns, then run the model. Because early results often diverge significantly from real-world observations, they “tune” these parameters to better match observed climate features. Different modeling teams use distinct assumptions and tuning strategies, leading to varied outcomes. Tuning is a necessary but delicate aspect of climate modeling, as it is for any complex system. Poor tuning can result in inaccurate simulations, while excessive tuning risks artificially steering results toward predetermined conclusions.

The spread of model representations of the current climate is very wide. One of the most basic indicators—Earth’s average surface temperature—varies by about 3°C across CMIP6 models prior to 1880 (Figure 5.1), narrows slightly until 2040 then diverges to over 4 °C. For comparison 20th century warming was only about 1.0°C. This variation suggests substantial differences among models’ physical processes.

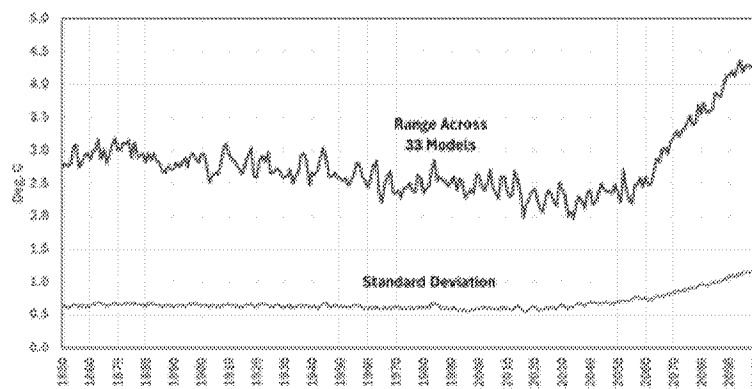


Figure 5.1 CMIP6 Average surface temperature range across 33 models and standard deviation using SSP5-85 scenario. Data from KNMI Climate Explorer website at [[HYPERLINK "https://climexp.knmi.nl/start.cgi?_blank"](https://climexp.knmi.nl/start.cgi?_blank)].

Beyond the models’ ability to reproduce features of today’s climate, the critical issue for society is how well they predict responses to subtle human influences, such as greenhouse gas emissions, aerosol cooling, and land-use changes. The most crucial aspect that models must capture correctly is “feedbacks.” These occur when climate changes either amplify or suppress further warming. In general, the modeled net effect of all feedbacks doubles or triples the direct warming impact of CO₂.

5.2 Surface warming

A straightforward test of a climate model's validity is its ability to reproduce historical warming in response to known past changes in climate drivers such as greenhouse gases. Figure 5.2 is reproduced from Scafetta (2023), that groups the latest-generation (CMIP6) climate models into low ECS (1.5 to 3.0 degrees C), medium ECS (3.0 to 4.5 degrees C) and high ECS (4.5 to 6.0 degrees C), and compares their post-1980 global average temperature simulation ranges to those of three surface temperature records and one satellite-based lower troposphere temperature data product.

The leftmost column shows that the low-ECS models track the post-1980 historical warming record reasonably well, but the middle and right columns show that the medium- and high-ECS models conspicuously over-predict the warming.

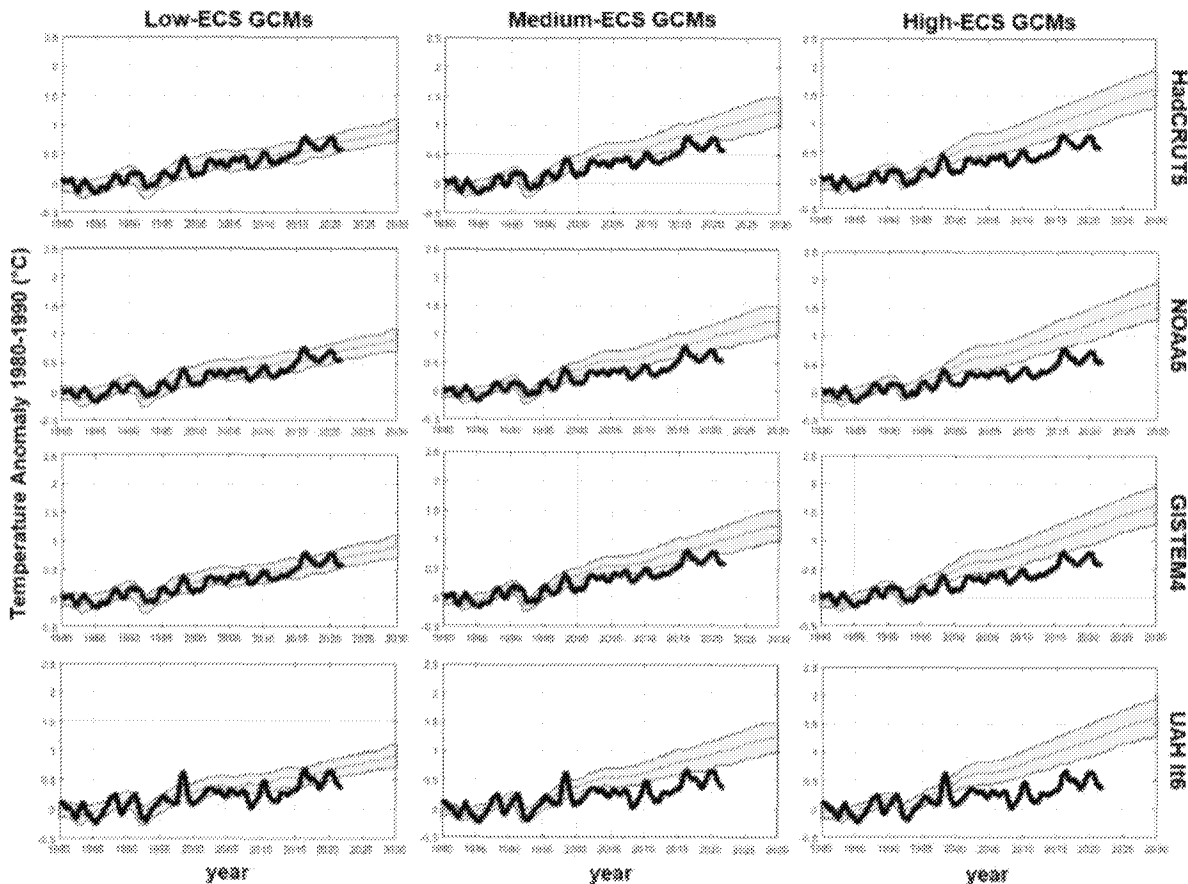


Figure 5.2: Model-Observational comparisons for Earth's surface warming. The columns correspond to model groups showing low-ECS (13 models) medium-ECS (11 models) and high-ECS (14 models), while the rows correspond to widely-used observed temperature records, the first three showing surface averages and the fourth showing the lower troposphere average. In each panel the yellow area denotes the mean and range ($\pm$ one standard deviation) of climate model simulations for that group. The thick black line shows the observed annual average temperature in the indicated record. Source: Scafetta (2023) Fig.2.

Spencer (2024) has also provided a useful summary of the Model-Observation mismatch by comparing trends in surface temperature data products with those in individual climate models, as summarized in Figure 5.3; most climate models show substantially more warming than the observations since 1979.

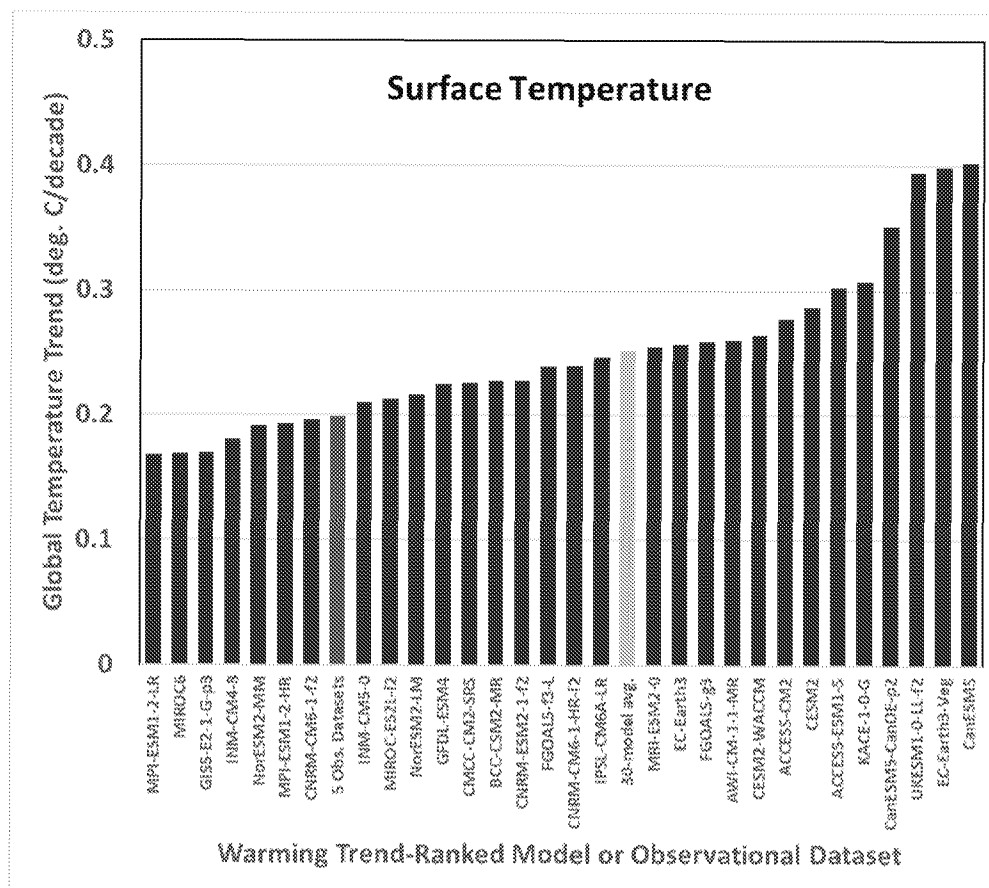


Figure 5.3 Global surface air temperature trends ($^{\circ}\text{C}/\text{decade}$), 1979-2024, from various CMIP6 climate models (red, 30-model average in orange); and the average of three thermometer datasets (HadCRUT5, NOAA Global Temp, and Berkeley 1 deg.) and two reanalysis datasets (ERA5 and NCEP/NCAR R1) in blue. Data source: <https://climexp.knmi.nl/start.cgi>.

5.3 Tropospheric warming

It has long been known that climate models on average overstate warming in the tropical troposphere. This region is an important test of climate models since this is where the signal of anthropogenic greenhouse warming emerges first and most strongly. Biases in tropospheric trends indicate model flaws in heat transfer processes that carry over to surface warming biases.

The discrepancy was flagged as a serious inconsistency in the first U.S. Climate Change Science Program report (Karl *et al.* 2006) and has been mentioned in every IPCC report since, but the discrepancy has gotten worse over time, and the bias is now global. McKittrick and Christy (2020) compared tropospheric warming trends in CMIP6 climate models to observed trends from satellites, weather balloons and reanalysis systems. Every model overpredicted the average observed warming trend over 1979-2014 in both Lower- and Mid-Troposphere layers, both globally and in the tropics. In most individual models the bias was statistically significant and on average across models it was highly significant.

Figure 5.4 presents the comparisons with data updated to 2024 (McKittrick and Christy 2025). The recent warm years moved the observed trend up slightly and widened the trend confidence intervals but the overall pattern remains the same: model bias is towards too much warming, in most cases the difference is statistically significant and on average the bias is statistically highly significant. McKittrick and Christy (2020) also showed that the bias is larger in high-ECS models, but even the models with lower average ECS predict too much warming. If future climate models were to realistically represent global tropospheric warming, they would likely be less sensitive than even the low-ECS members of the CMIP6 ensemble.

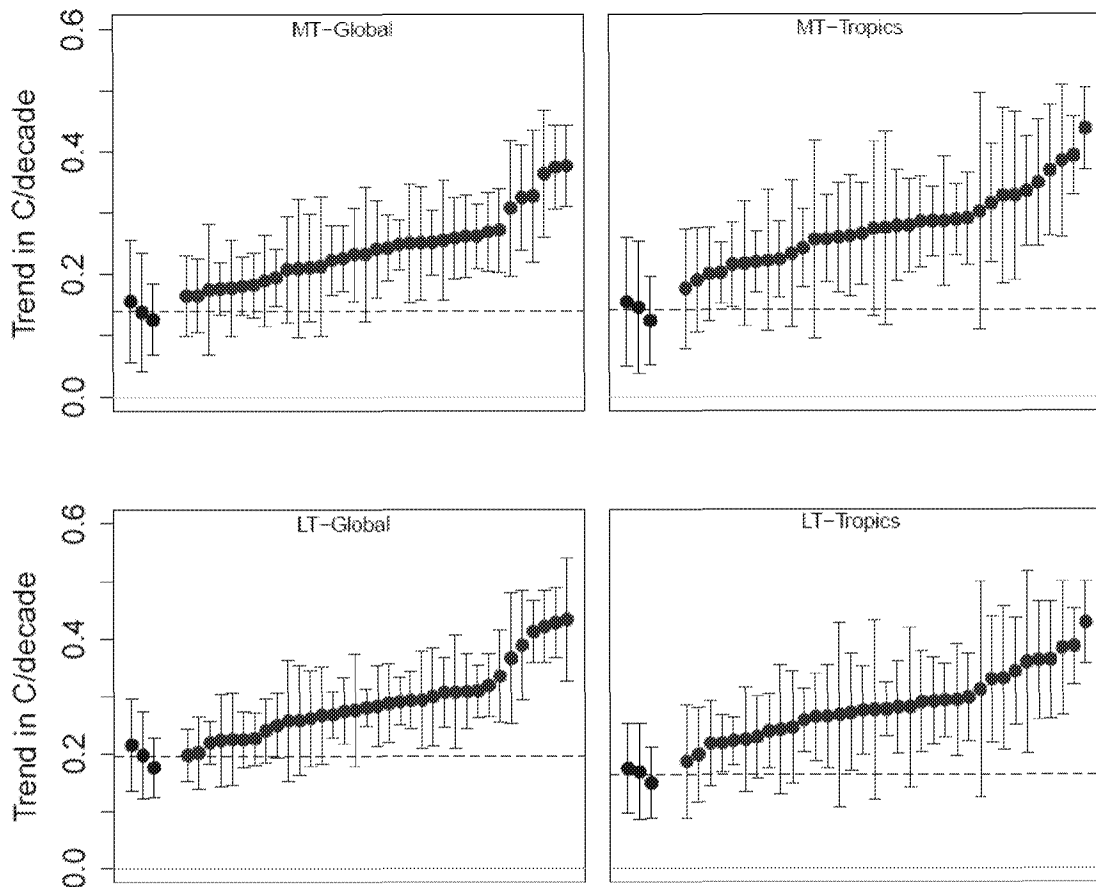


Figure 5.4: Observed versus CMIP6 modeled warming trends ($^{\circ}\text{C}/\text{decade}$ 1979-2024) in the global and tropical lower (LT) and mid-troposphere (MT) using the methodology of McKittrick and Christy (2020) on data updated from 2014 to 2024. Blue dots: warming trends with 95 percent confidence intervals for 3 data products (radiosondes, reanalysis, and satellites). Blue dashed line: warming trend average for 3 observed series. Red dots: modeled warming trends with 95 percent confidence intervals in 35 models arranged lowest to highest.

As mentioned previously, the IPCC has long acknowledged the model-observation mismatch. For example, AR6 pp. 443-444 offers this on the tropical troposphere (it does not address the global comparison):

Several studies since AR5 have continued to demonstrate an inconsistency between simulated and observed temperature trends in the tropical troposphere, with models simulating more warming than observations (Mitchell *et al.*, 2013, 2020; Santer *et al.*, 2017a, b; McKittrick and Christy, 2018; Po-Chedley *et al.*, 2021)... Over the 1979–2014 period, models are more consistent with observations in the lower troposphere, and least consistent in the upper troposphere around 200 hPa, where biases exceed 0.1°C per decade. Several studies using CMIP6 models suggest that differences in climate sensitivity may be an important factor contributing to the discrepancy between the simulated and observed tropospheric temperature trends (McKittrick and Christy, 2020; Po-Chedley *et al.*, 2021), though it is difficult to deconvolve the influence of climate sensitivity, changes in aerosol forcing and internal variability in contributing to tropospheric warming biases (Po-Chedley *et al.*, 2021). Another study found that the absence of a hypothesized negative tropical cloud feedback could explain half of the upper troposphere warming bias in one model (Mauritsen and Stevens, 2015).

... In summary, studies continue to find that CMIP5 and CMIP6 model simulations warm more than observations in the tropical mid- and upper-troposphere over the 1979–2014 period (Mitchell *et al.*, 2013, 2020; Santer *et al.*, 2017a, b; Suárez-Gutiérrez *et al.*, 2017; McKittrick and Christy, 2018), and that overestimated surface warming is partially responsible (Mitchell *et al.*, 2013; Po-Chedley *et al.*, 2021). Hence, we assess with *medium confidence* that CMIP5 and CMIP6 models continue to overestimate observed warming in the upper tropical troposphere over the 1979–2014 period by at least 0.1°C per decade,

Notably, despite the accumulation of evidence of excess model warming the IPCC assigns only *medium confidence* to the existence of a warming bias.

5.4 Vertical temperature profile mismatch

Another important model-observational discrepancy is the excess amplification with altitude found in climate models. The comparison was in AR5 Chapter 10, although only in the online supplement (Figure 10.SM.1) and only in a figure whose formatting obscured the point. Figure 10.SM.1 is not referenced in the main IPCC report nor in any summary so readers would not have been aware of it. Although not apparent at first glance, it shows that the 1979–2010 warming in the lower troposphere is so small as to be consistent with no GHG forcing at all and is inconsistent with the model runs that do have GHG forcing. In Figure 5.6 we adapt IPCC AR5 Figure 10.SM.1 to draw out this critical point.

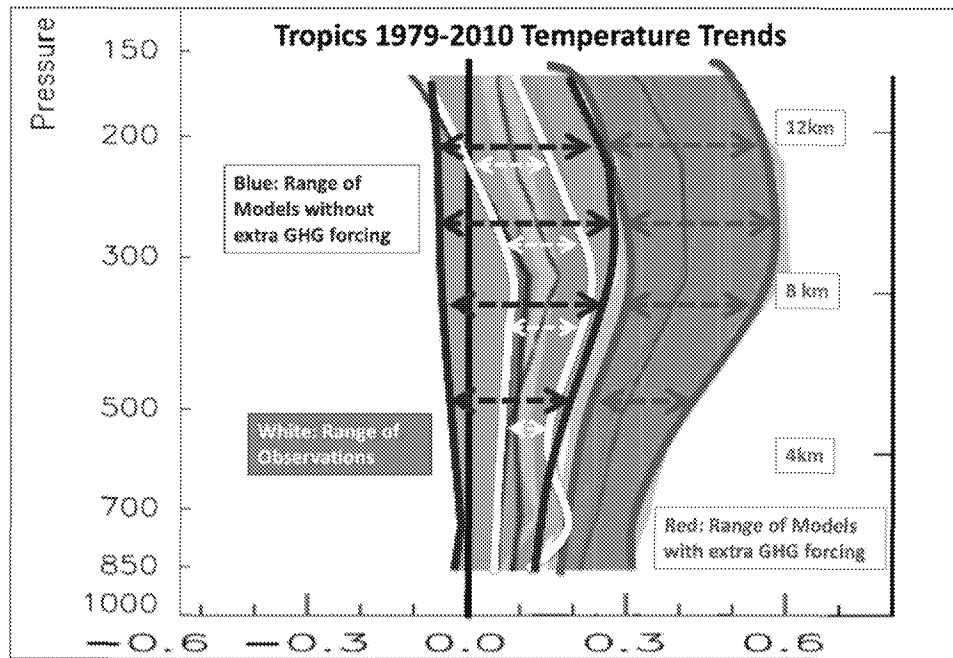


Figure 5.5: Vertical warming pattern for tropics (20S to 20N). Horizontal axis: °C/decade. Source: Annotated version of IPCC AR5 Figure 10.SM.1

Figure 5.5 compares model and observational temperature trends by altitude between 20S and 20N (the tropics). In this region where the models say the warming should be strongest, the observations (shown here in white) lie within the blue “No CO₂” band and entirely outside the “with CO₂” red envelope. This means that in the entire tropical atmospheric column from the surface to the base of the stratosphere, observed warming trends are so small as to be consistent with the output of models that have no anthropogenic CO₂, and inconsistent with the entire envelope of warming trends generated by models forced with increased CO₂.

A similar comparison is shown in Christy and McNider (2017), an updated version of which (covering 1979-2024) is reproduced as Figure 5.6. Modeled temperature trends exceed observations from the surface through the top of the troposphere, with observed trends below the entire model range at most pressure levels. Also shown in Figure 5.6 is the tropical tropospheric temperature (TTT) average from three satellite data products (NOAA, UAH and RSS) compared to the same layer average from climate models for 1979-2024. Again, the observed trends lie below the entire model range.

The wide range of choices made by modelers to characterize the physical processes in the models (see Box: Climate Modeling in Section 5.1 above) is seen by the large spread of trends in the middle troposphere, ± 40 percent about the median (Figure 5.6). This vividly illustrates the uncertainties in attempts to model (parameterize) a complex system involving turbulence, moist thermodynamics, and energy fluxes over the full range of the tropical atmosphere’s time and space scales.

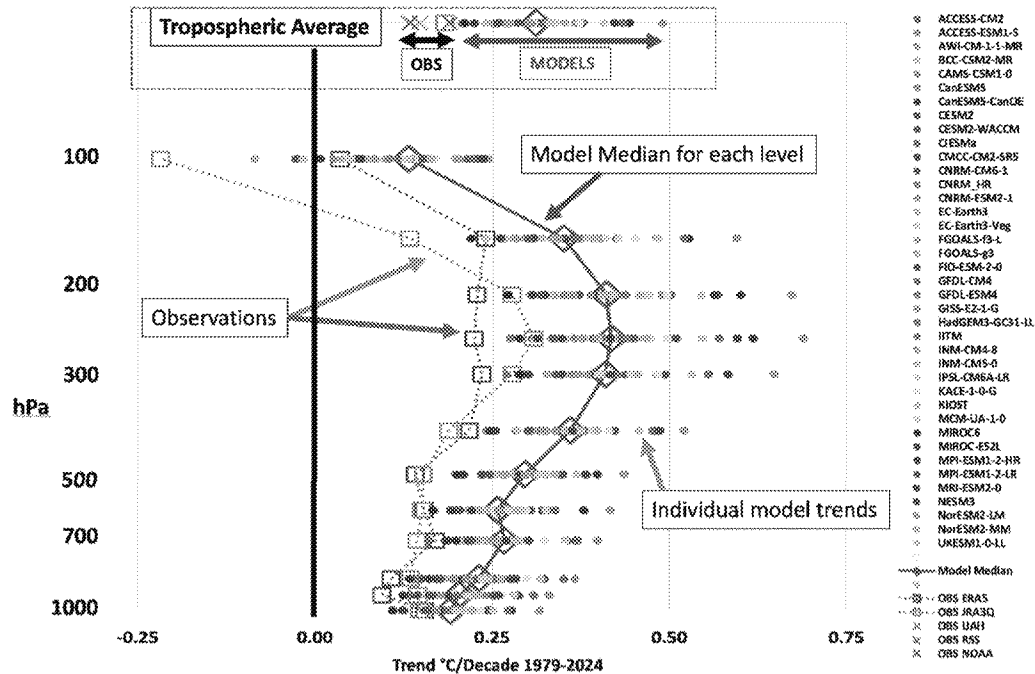


Figure 5.6: Modeled versus observed warming, tropical troposphere. Source: updated from Christy and McNider (2017) including data through 2024 and CMIP6 model outputs. Red line: model average. Green and blue lines: observational series (reanalysis).

This discrepancy has been the source of much controversy, with some arguing that even if there is very little observed warming aloft in the tropics, a “hotspot” still exists in the sense that the warming aloft is greater than at the surface (Santer et al. 2008). But there is good evidence that models also exaggerate the amplification rate. Klotzbach *et al.* (2009), showed that models project greater amplification with altitude than is observed. This result was subsequently confirmed by detailed time series analysis (Vogelsang and Nawaz 2016) which found that the model-observational difference is statistically significant.

The atmosphere’s temperature profile is a case where models are not merely uncertain but also show a common warming bias relative to observations. This suggests that they misrepresent certain fundamental feedback processes.

The IPCC AR6 did not assess this issue.

5.5 Stratospheric cooling

An important element of the expected general “fingerprint” of anthropogenic climate change is simultaneous warming of the troposphere and cooling of the stratosphere. The latter feature is also influenced by ozone depletion and recovery. AR6 acknowledged that cooling had been observed but only until the year 2000. The stratosphere has shown some warming since, contrary to model projections.

AR6 WG1 Ch 2 pp. 327-9 states:

Temperatures averaged through the full lower stratosphere (roughly 10–25 km) have decreased over 1980–2019 in all data products, with the bulk of the decrease prior to 2000. The decrease holds even if the influence of the El Chichon (1982) and Pinatubo (1991) volcanic eruptions on the trend, found by Steiner *et al.* (2020a) to have increased the 1979–2018 cooling trend by 0.06°C per decade, is removed. Most datasets show no significant or only marginally significant trends over 2000–2019, and the results of Philipona *et al.* (2018) show weak increases over 2000–2015 in the very lowermost stratosphere sampled by radiosondes....

It is virtually certain that the lower stratosphere has cooled since the mid-20th century. However, most datasets show that lower stratospheric temperatures have stabilized since the mid-1990s with no significant change over the last 20 years. It is likely that middle and upper stratospheric temperatures have decreased since 1980, but there is low confidence in the magnitude.

The cited source, Philipona *et al.* (2018), in an article entitled “Radiosondes Show That After Decades of Cooling, the Lower Stratosphere Is Now Warming”, stated:

In response to continued greenhouse gas increases and stratospheric ozone depletion, climate models project continued tropospheric warming and stratospheric cooling over the coming decades. Global average satellite observations of lower stratospheric temperatures exhibit no significant trends since the turn of the century. In contrast, an analysis of vertically resolved radiosonde measurements from 60 stations shows an increase of lower stratospheric temperature since the turn of the century at altitudes between 15 and 30 km and over most continents. Trend estimates are somewhat sensitive to homogeneity assessment choices, but all investigated radiosonde data sets suggest a change from late twentieth century cooling to early 21st century warming in the lower stratosphere.

Santer *et al.* (2023) use updated data to show that a cooling trend has not re-emerged in the lower stratosphere.

A combination of tropospheric warming and stratospheric cooling is a commonly cited “fingerprint” of anthropogenic climate change. Stratospheric warming since 2000 coincides with continued surface and tropospheric warming, a pattern that is not found in climate model simulations and is not apparently consistent with the anthropogenic fingerprint.

5.6 Snow cover mismatch

Data compiled by the Rutgers University Snow Lab show that Northern Hemisphere winter snow cover is not decreasing (Figure 5.7); if anything, it shows an increasing trend.

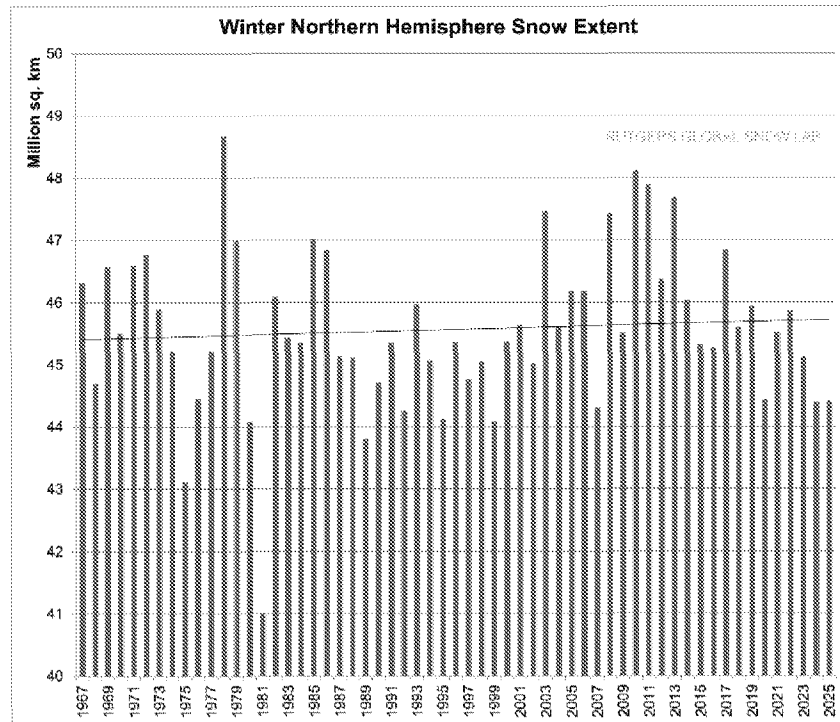


Figure 5.7: Northern Hemisphere Winter Snow extent.

Source: [[HYPERLINK](https://climate.rutgers.edu/snowcover/chart_seasonal.php?ui_set=nhland&ui_season=1)]

"https://climate.rutgers.edu/snowcover/chart_seasonal.php?ui_set=nhland&ui_season=1"]
(accessed May 27, 2025)

Yet models project declining Northern Hemisphere snow cover in a warming climate, as described by Connolly *et al.* (2019).

The climate models were found to poorly explain the observed trends [in Northern Hemisphere snow cover]. While the models suggest snow cover should have steadily decreased for all four seasons, only spring and summer exhibited a long-term decrease, and the pattern of the observed decreases for these seasons was quite different from the modelled predictions. Moreover, the observed trends for autumn and winter suggest a long-term increase, although these trends were not statistically significant.

AR6 largely confines its discussion of changing Northern Hemisphere snow cover extent (SCE) to the Spring season, for which models and observations agree on a downward trend. Regarding Winter changes it remarks as follows (AR6 WGI Ch. 2 p. 344):

Assessment of SCE trends in the NH since 1978 indicates that for the October to February period there is substantial uncertainty in trends with the sign dependent on the observational product. Analysis using the NOAA Climate Data Record shows an increase in October to February SCE (Hernández-Henríquez *et al.*, 2015; Kunkel *et al.*, 2016) while analyses based on satellite borne optical sensors (Hori *et al.*, 2017) or multi-observation products (Mudryk *et al.*, 2020) show a negative trend for all seasons.

AR6 WGI Chapter 9 (p. 1284) points out that the NOAA Climate Data Record showing increased fall and winter SCE is inconsistent with land-based observations and model-based data sets. It notes that using optical satellite imagery to infer SCE in winter is challenging due to cloud cover and decreased solar illumination in winter months. Focusing on the Pacific Coast states (CA, OR and WA) cold season mountain snowfall that melts in spring and summer provides a substantial portion of warm season water resources. A comprehensive reconstruction of snowfall for the main source regions (Cascades and Sierra Nevada Mountains) indicates no significant trends in annual totals since the late 19th century (Christy 2022).

In summary, the up-to-date Rutgers SCE database indicates a mismatch between models and observations. Further work to reconcile conflicting trends in observational data sets is required.

5.7 Hemispheric symmetry of the planetary albedo

The planetary albedo is the fraction of incoming solar radiation reflected by the Earth back to space. It is an important element of the radiative energy balance and influences whether the planet will warm or cool over time. The planetary albedo is typically estimated at around 0.30; variations on the scale of 0.01 correspond to changes in solar forcing of about 3 W/m^2 , an amount larger than current anthropogenic forcing. It has long been noted that models disagree with each other and with observations on the value of the global planetary albedo (Stephens *et al.* 2015).

An intriguing property of the Earth's albedo is that, on average, the Northern Hemisphere (NH) and Southern Hemisphere (SH) have had nearly the same albedo, at least throughout the fifty-year satellite record (Stephens *et al.*, 2015). This symmetry is surprising, because the SH has much more ocean than land. Since ocean is less reflective than land, the NH should have higher albedo. Clouds (which are highly reflective) are more common in the NH and so compensate the surface albedo imbalances of the two hemispheres. Datseris and Stephens (2021) show that this cloud compensation comes from the extra-tropical storm tracks of the SH, which are cloudier than those of the NH. While the mechanism for this hemispheric symmetry is unclear, it likely operates on large temporal and spatial scales.

The hemispheric symmetry of the albedo is a simple gross metric for climate models. Rugenstein and Hakuba (2023) defined that metric as the difference between the NH and SH annual mean albedos, expressed as Wm^{-2} of the reflected sunlight, and compiled it for the CMIP6 climate models, as shown in Figure 5.8. Most of the CMIP6 models do not reproduce the small observed asymmetry (about 0.1 Wm^{-2}) and even disagree as to which hemisphere is more reflective. Moreover, the magnitude of the asymmetry ranges up to 5 Wm^{-2} in some models, twice as large as the current anthropogenic forcing (about 2.7 Wm^{-2}).

The significance of unphysical albedo asymmetries in the climate models is not yet fully known. However, other model studies suggest that interhemispheric changes in albedo can alter poleward heat fluxes, meridional temperature gradients, storminess, and differences in hemispheric ocean heat storage. The discrepancy between models and observations further raises issues regarding cloud feedback processes and so more generally diminishes confidence in model projections of the future climate.

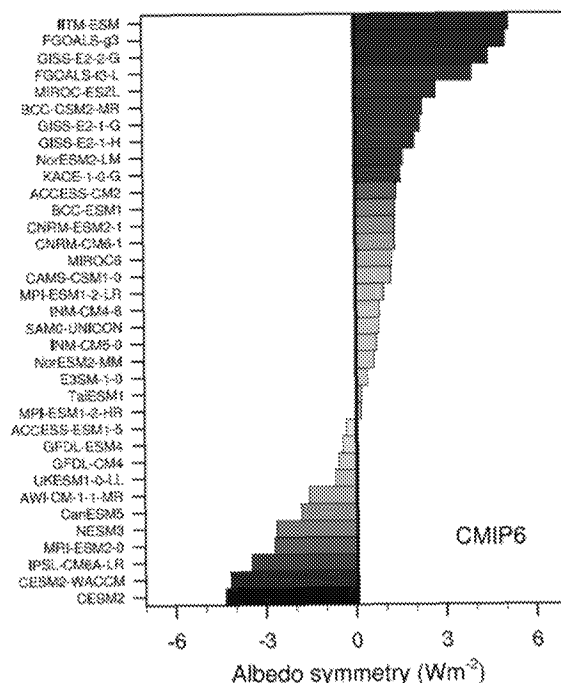


Figure 5.8. Differences in 20-year average reflectivity (albedo) between the Northern and Southern Hemispheres for CMIP6 models used in the most recent IPCC assessment (colored bars). The very small observed difference is indicated by the vertical black line. From Rugenstein and Hakuba (2023).

5.8 U.S. Corn Belt

One of the largest discrepancies between models and observations is in the U.S. Corn Belt, a region of particular importance to global food production. Figure 5.9 shows the warming trends for summertime (June, July, August) for the 12-state Corn Belt (IN, IA, IL, ND, SD, MO, MN, WI, MI, OH, KS, NE) during 1973-2022. All 36 climate models (red) warm far too rapidly compared to observations (blue).

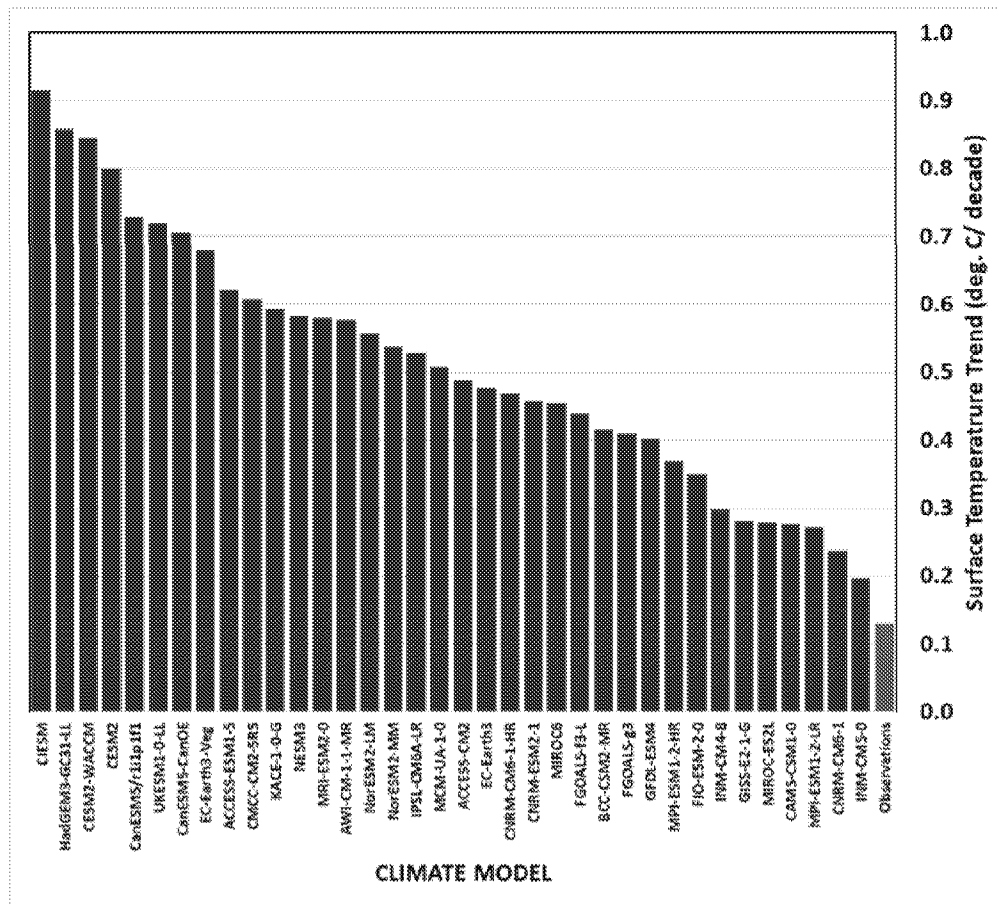


Figure 5.9: Modeled versus observed warming trends in the U.S. Corn Belt, 1973-2022.

As discussed in Chapter 9, the anticipated negative effects of increasing temperatures on U.S. corn yields have not materialized, in contrast to widely publicized studies proclaiming that theoretical future impacts are already being experienced (*e.g.*, Seager *et al.*, 2018).

The IPCC acknowledges limitations in the accuracy of regional climate model outputs. This example shows that users need to assess model projections carefully on a case-by-case basis since local biases might be sufficiently large that the models are simply not fit for purpose. As has recently been noted by two leaders of the modeling community (emphasis added)

... for many key applications that require regional climate model output or for assessing large-scale changes from small scale processes, we believe that **the current generation of models is not fit for purpose**. (Palmer and Stevens 2019)

To summarize:

- Climate models show warming biases in many aspects of their reproduction of the past few decades.
- They produce too much warming at the surface (except in the models with lowest ECS), too much warming in the lower-and mid-troposphere and too much amplification of warming aloft

- They also produce too much stratospheric cooling, too much snow loss, and too much warming in the U.S. Corn Belt.
- The hemispheric albedo difference in individual climate models ranges widely in sign and magnitude compared to observations. The range in W/m^2 is three times larger than the direct anthropogenic forcing of CO_2 .
- The IPCC has acknowledged some of these issues but not others.

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6 EXTREME WEATHER

Chapter Summary

Most types of extreme weather exhibit no statistically significant long-term trends over the available historical record. While there has been an increase in hot days in the U.S. since the 1950s, a point emphasized by AR6, numbers are still low relative to the 1920s and 1930s. Extreme convective storms, hurricanes, tornadoes, floods and droughts exhibit considerable natural variability, but long-term increases are not detected. Some increases in extreme precipitation events can be detected in some regions over short intervals but the trends do not persist over long periods and at the regional scale. Wildfires are not more common in the U.S. than they were in the 1980s. Burned area increased from the 1960s to the early 2000's, however it is low compared to the estimated natural baseline level. U.S. wildfire activity is strongly affected by forest management practices.

6.1 Introduction

High impact weather extremes, usually related to temperature, precipitation and/or high wind, can disrupt infrastructure and therefore endanger human health and wellbeing. The issue is not whether extremes occur. Rather, it is whether there are long-term (decadal scale) changes in the frequency or character of extremes (“detection”), as well as the extent to which such changes and the attendant changes in hazards are caused by anthropogenic emissions of greenhouse gases (“attribution”; e.g., AR6 Seneviratne *et al.* 2021).

Process-based understanding and simple thermodynamic arguments have been invoked to assert that warming is worsening extreme weather events. However, it is naïve to assume that any recent extreme event is caused by human influences on the climate. Climate is about the statistical properties of weather over decades, not single events. Further, there are only about 130 years of reliable observational records that can be analyzed statistically. That brief interval does not begin to contain all the extreme events that the climate system can create on its own. Over geologic time the climate system has generated an (essentially) infinite variety of weather patterns and extremes that humans have never observed and thus are absent from the databases used to determine extreme statistics [see *Perils of short data records* below]. For that reason, attributing an extreme event unprecedented in the record requires assumptions about the magnitude of natural variations.

This chapter is concerned with detection of trends in extreme weather, while Chapter 8 considers causal attribution, with Section 8.4 specifically addressing extreme weather. If no trend is detected, then clearly there is no basis for attribution. But even where a trend is observed, attribution to human-caused warming does not necessarily follow.

This is especially true for precipitation events. The hydrological literature has long noted the presence of long, slow and irregular oscillations in rainfall data (Hurst 1951, Cohn and Lins 2005, Markonis and Koutsoyiannis 2016). The characteristics of these natural patterns require long records to accurately estimate variability. Analysis of records that are short relative to the scale of natural variability will tend to misrepresent trends, therefore overstating the significance of apparent trends and underestimating the likelihood of extreme events (Cohn and Lins 2005).

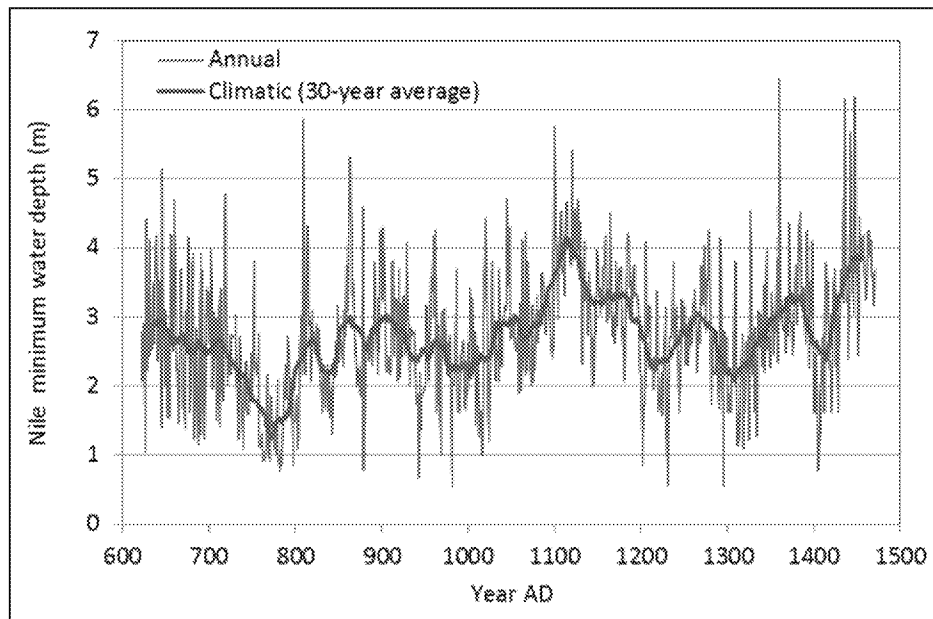


Figure 6.1.1: The annual minimum depth of the Nile River near Cairo over more than 650 years from 622 to 1284 A.D. The data, measured in meters, shows a characteristic pattern of year-to-year fluctuations around longer-term trends. Data from Koutsoyiannis (2013)

A good example of this is the eight-century long record of the annual minimum height of the Nile River observed at Roda Island in Cairo shown in Figure 6.1.1. The Nile River is fed by precipitation over a 4 million square mile drainage basin, an area equal to about one third of CONUS. Since human influences on the global climate were negligible long before the 20th century, the century-scale variability of the thirty-year average is entirely natural; Egyptians of the seventh and eight centuries would have been incorrect to assume that the worsening drought during that time was the “new normal.”

With these caveats in mind, we examine the evidence for changes in selected weather and climate extremes. A recurring theme is the wide gap between public perceptions and scientific evidence. It has become routine in media coverage, government and private sector discussions, and even in some academic literature to make generalized assertions that extreme weather of all types is getting worse due to GHGs and “climate change.” Yet expert assessments typically have not drawn such sweeping conclusions and instead have emphasized the difficulty both of identifying specific trends and establishing a causal connection with anthropogenic forcing.

In the sections to follow we provide excerpts from various IPCC and NCA assessment reports denoting the sources as follows:

SREX: The IPCC Special Report on Managing the Risks of Extreme Events and Disasters to Advance Climate Change Adaptation (2012)

AR6: The IPCC Sixth Assessment Report Working Group 1 (2021).

NCA4: The U.S. Climate Science Special Report of the Fourth National Climate Assessment (2017) Volume I.

NCA5: Fifth National Climate Assessment (2023).

In the excerpts, *italics* are in the original whereas **boldface** emphasis has been added.

Additionally we use standard government sources to provide information through 2024 wherever possible.

6.2 Hurricanes and tropical cyclones

AR6 provides the following assessment of tropical cyclones (TCs; used here as a synonym for hurricanes):

AR6: There is *low confidence* in most reported long-term (multidecadal to centennial) trends in TC frequency or intensity-based metrics due to changes in the technology used to collect the best-track data. (IPCC, 2021 p. 1585)

AR6: It is *likely* that the global proportion of major (Category 3–5) tropical cyclone occurrence has increased over the last four decades . . . There is *low confidence* in long-term (multi-decadal to centennial) trends in the frequency of all-category tropical cyclones. (IPCC, 2023 SPM p. 9)

AR6: A subset of the best-track data corresponding to hurricanes that have directly impacted the United States since 1900 is considered to be reliable, and shows **no trend in the frequency of U.S. landfall events**. (IPCC 2021 p. 1585)

Since 1980, when satellite observations first fully covered the global oceans, we have confidence in the numbers of total global hurricanes and major hurricanes (Category 3 and higher). Figure 6.2.1 shows that on average, each year there are about 50 hurricanes, with about 25 reaching major hurricane status (Maue, 2025). There is substantial year-to-year and decadal variability, a weak decrease in the number of hurricanes, and a slight but insignificant increase in the number of major hurricanes. These two trends combine to create an overall increase in the proportion of major hurricanes.

Global hurricane statistics are dominated by the Northwest Pacific Ocean, which accounts for ~35 percent of total global hurricanes, whereas the Atlantic accounts for ~15 percent of global hurricanes (Colorado State University, 2025). Data in the Atlantic basin extends further back than in the other ocean basins and is most relevant to U.S. policy makers.

Global Major Hurricane Frequency -- 12 month running sums -- @RyanMaue

Updated March 10, 2025

Last 30-years, annual: 45 H | 24 M

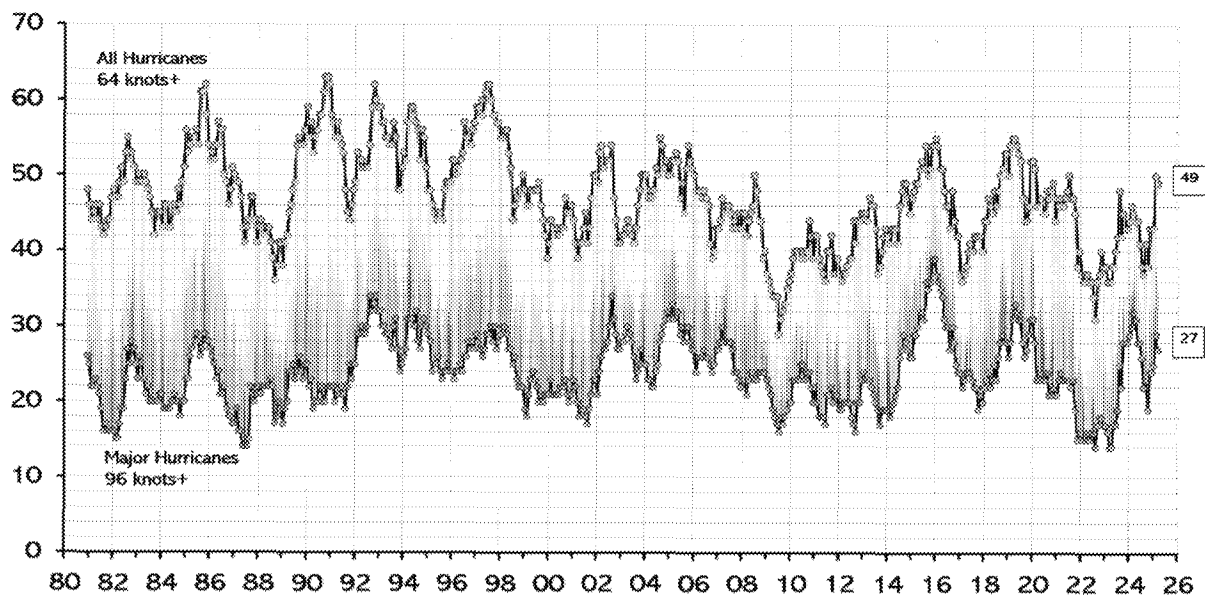


Figure 6.2.1: Global frequency of hurricanes and major hurricanes since 1980. Source: Updated from Maue 2011.

Figure 6.2.2 shows the frequency of Atlantic hurricanes and major hurricanes (Category 3 and higher) back to 1920. Data prior to 1965 (the onset of satellite observations in the Atlantic, shaded in blue) shows some undercounting, with data prior to 1920 showing substantial undercounting (Vecchi and Knutson, 2011). All measures of Atlantic hurricane activity show a significant increase since 1970. However, the period from 1971-1994 saw exceptionally low activity, with high activity (comparable to the past two decades, even with undercounting) also observed during the 1950's and 1960's, and even in the 1930's.

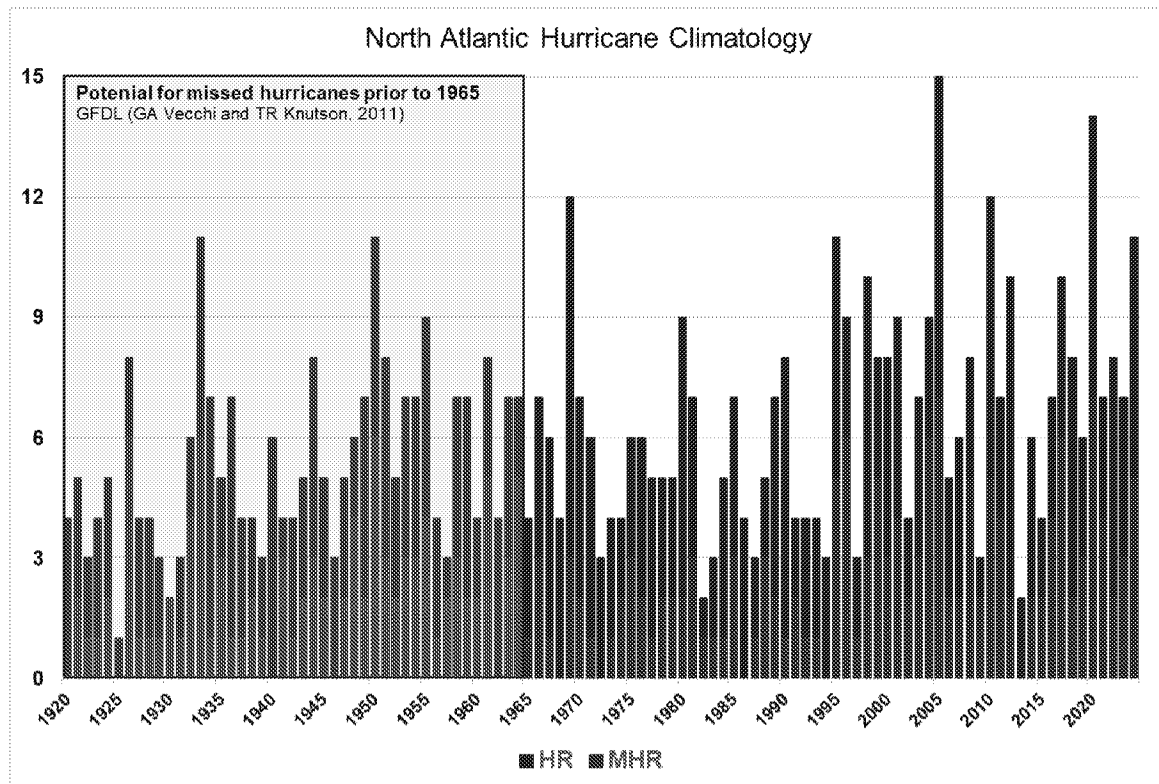


Figure 6.2.2: Atlantic frequency of hurricanes (HR) and major hurricanes (MHR) since 1920. Source National Hurricane Center (2024)

Figure 6.2.2 shows that Atlantic hurricanes vary strongly on decadal and multidecadal time scales. These variations are associated primarily with the Atlantic Multidecadal Oscillation (AMO), which is manifest in basin-wide sea surface temperature and sea level pressure fluctuations connected to large scale ocean circulation patterns. The AMO was in its warm phase during 1926-1970 and 1995-present, but in its cool phase during 1971-1994. It has its greatest impact on the number of major hurricanes (Category 3+), which Goldenberg *et al.* (2001) associated with above normal SSTs and decreased vertical shear in the AMO warm phase (see also Bell and Chelliah, 2006; Klotzbach *et al.*, 2018).

Klotzbach *et al.* (2018) conducted a comprehensive evaluation of the landfalling hurricane data for the Continental U.S. since 1900. Figure 6.2.3 updates their analysis through 2024. While the largest numbers of landfalling hurricanes are from 2004, 2005 and 2020, there is no statistically significant trend since 1920. Figure 6.2.3 also shows the time series for major hurricane landfalls (Category 3-5). The largest year in the record is 2005, with 4 major hurricane landfalls. However, following 2005 there were no major hurricanes striking the U.S through 2016, the longest such period since 1920.

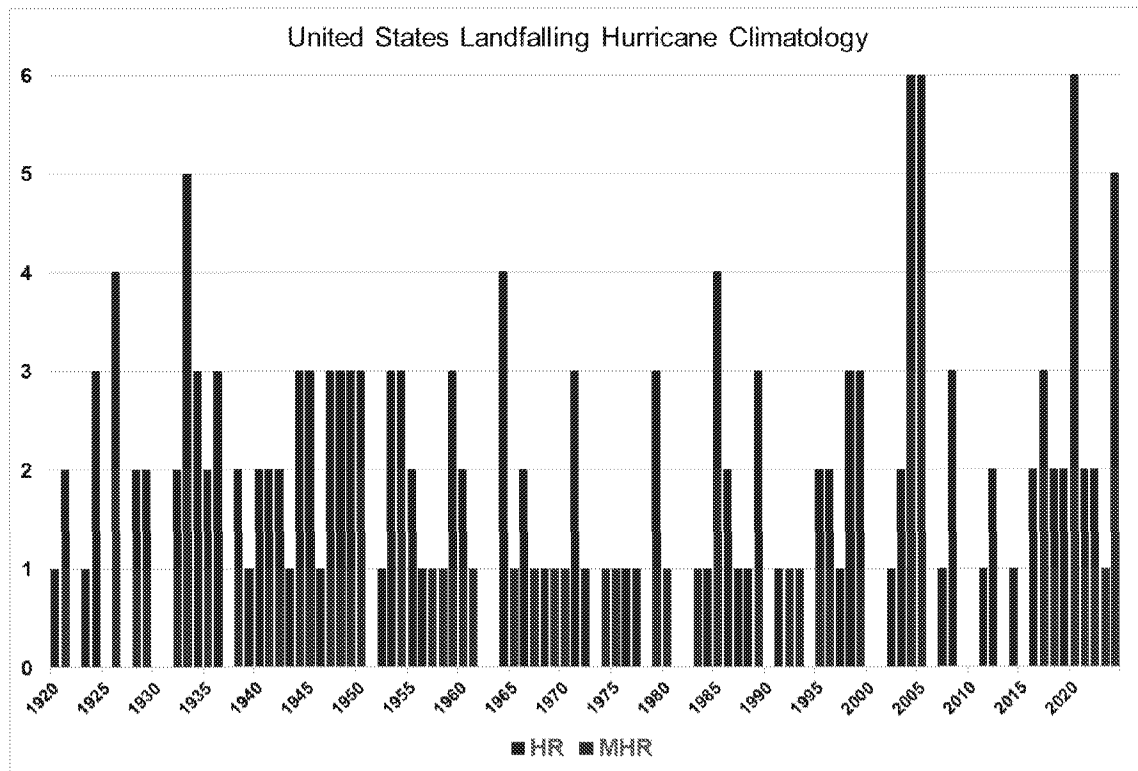


Figure 6.2.3: U.S. landfalling frequency of hurricanes (HR) and major hurricanes (MHR) since 1920. Source NOAA HRD(a) (2024)

Figure 6.2.3 shows substantial interannual to multidecadal variability in U.S. landfall activity. Klotzbach *et al.* (2018) examined how the landfall counts vary with ENSO (El Niño versus La Niña) and the warm versus cold phases of the Atlantic Multidecadal Oscillation (AMO).

Villarini *et al.* (2012) provide an analysis of U.S. hurricane landfalls back to 1878. While it is possible that some landfalls were missed in the late 19th century owing to sparsely populated regions on the Gulf Coast, it is remarkable that the highest year in the entire record is 1886, with 7 hurricane landfalls, when human influences on the climate were much smaller than they are today.

Table 6.2.1 shows the 10 strongest hurricanes (plus ties) to make U.S. landfall. Of the hurricanes that have made landfall with sustained winds greater than 150 mph, only one has occurred in the 21st century.

In summary, analyses of both global and regional variability and trends of hurricane activity provide the basis for detecting changes and understanding their causes. The relatively short historical record of hurricane activity, and the even shorter record from the satellite era, is not sufficient to assess whether recent hurricane activity is unusual relative to the background natural variability. Atlantic hurricane processes are influenced substantially by the natural modes of ocean circulation variability in the Atlantic, notably the Atlantic Multidecadal Oscillation. While it has long been hypothesized that a rising global sea surface temperature would cause an increase in hurricane intensity, identification of any significant trend in the hurricane data is hampered by a short data record and substantial natural variability.

Process-based understanding also suggests that storm surges and rainfall from hurricanes should be increasing with warmer temperatures. However, the relatively small number of hurricanes with varying landfall locations and the complex dynamics associated with each storm preclude meaningful detection of change.

Rank	Year	Landfall Wind (MPH)	Name
1	1935	185	"Labor Day"
2	1969	175	Camille
3	1992	165	Andrew
4	2018	160	Michael
5	1856	150	"Last Island"
5	1886	150	"Indianola"
5	1919	150	-----
5	1932	150	"Freeport"
5	2004	150	Charley
5	2020	150	Laura
5	2021	150	Ida
5	2022	150	Ian

Table 6.2.1 Strongest hurricanes to make landfall along the U.S. coast. Source (NOAA HRD(b), 2024)

6.3 Temperature extremes

The AR6 assessment focused on the period after 1950 and reported increasing trends in heatwave frequency and intensity. However, NCA4 noted that heatwave activity in the U.S. reached a peak in the 1930s (Figure 6.3.1).

AR6: It is *virtually certain* that hot extremes (including heatwaves) have become more frequent and more intense across most land regions since the 1950s, while cold extremes (including cold waves) have become less frequent and less severe (SPM, A3.1)

AR6: In North America, there is very robust evidence for a *very likely* increase in the intensity and frequency of hot extremes and decrease in the intensity and frequency of cold extremes for the whole continent, though there are substantial spatial and seasonal variations in the trends. Minimum temperatures display warming consistently across the continent, while there are more contrasting trends in the annual maximum daily temperatures in parts of the USA. (Chapter 11, p 1550)

NCA4: Changes in warm extremes are more nuanced than changes in cold extremes. For instance, the warmest daily temperature of the year **increased** in some parts of the West over the past century, but there were **decreases** in almost all locations east of the Rocky Mountains. In fact, all eastern regions experienced a net **decrease**, most notably the Midwest (about 2.2°F [1.2°C]) and the Southeast (roughly 1.5°F [0.8°C]). (pp. 190-191)

NCA4: Since the mid-1960s, there has been only a very slight increase in the warmest daily temperature of the year (amidst large interannual variability). Heat waves (6-day periods with a maximum temperature above the 90th percentile for 1961–1990) increased in frequency until the mid-1930s, became considerably less common through the mid-1960s, and increased in frequency again thereafter. As with warm daily temperatures, **heat wave magnitude reached a maximum in the 1930s.** (pp. 190-191)

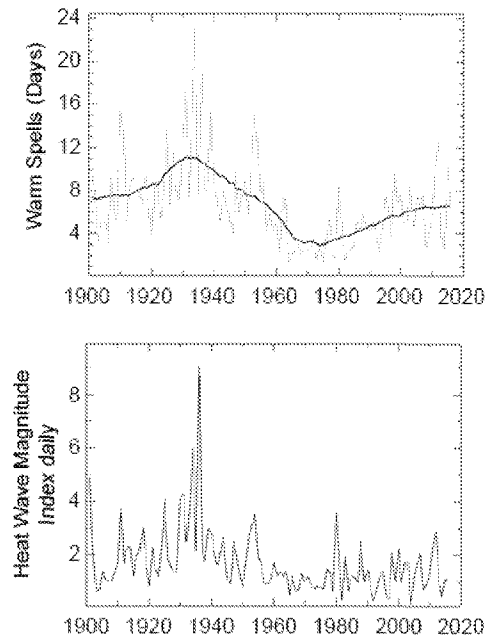


Figure 6.3.1: U.S. Heat waves since 1900. Source: NCA4 Figure 6.4

6.3.1 *Temperatures in the U.S. are becoming less extreme*

Daily maximum temperatures in the warm season (Tmax, May-Sep) and daily minimum temperatures in the cold season (Tmin, Dec-Mar) are available beginning in Dec 1898 (126 years). The dataset consists of 1,211 CONUS stations designated as United States Historical Climate Network or USHCN stations (see Figure 6.3.2; Quinlan *et al.* 1987, Karl *et al.* 1990). These stations were selected by NOAA as having the fewest problematic issues with gaps, station moves, and instrument changes. Where gaps still exist, nearby stations (bias-corrected) were merged so that the median volume of data available for a station is 98%. Although there are certainly errors in the dataset, including unresolved spurious warming due to UHI effects that especially bias Tmin records, this data set is sufficiently accurate for assessing trends in Tmax heat extremes.

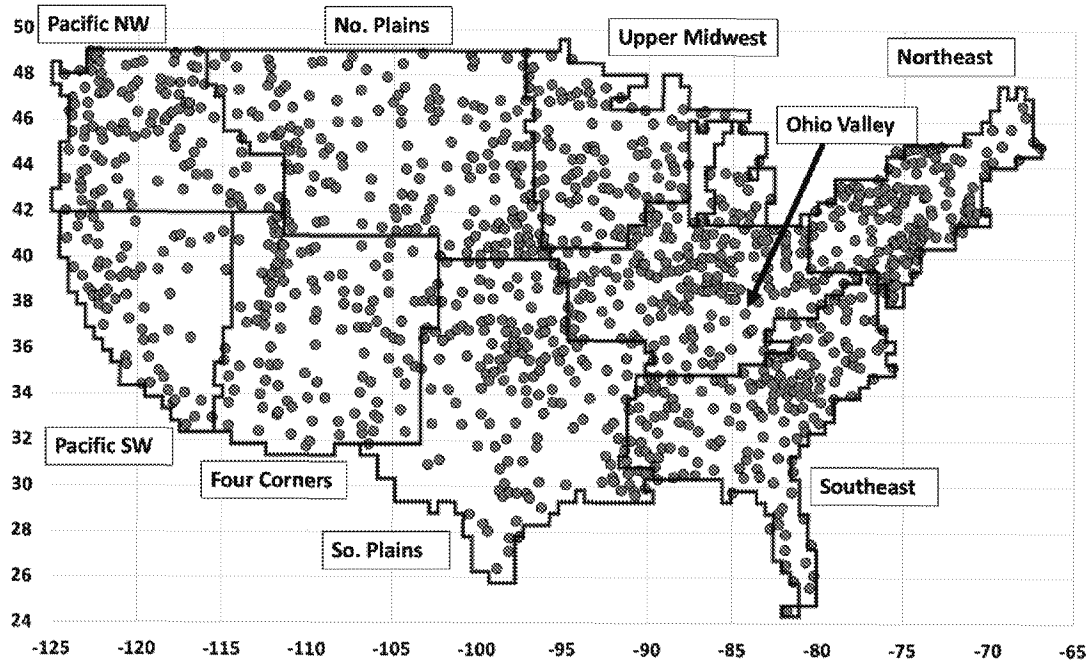


Figure 6.3.2 Locations of USHCN temperature stations. Source: USHCN.

We begin with the question of whether the occurrence of daily record high or low temperatures has changed since Dec 1898. Each warm-season has 153 days (1 May to 30 Sep) and each cold-season has 122 days (1 Dec to 31 Mar). For each station and day, we calculated the year in which the record highest (lowest) temperature occurred. With 126 years of observations, if there were no temperature trends over time, the expected number of records for Tmax would be 1.21 ($=153/126$) per station per year and for Tmin 0.96 ($=122/126$) per station per year.

Figure 6.3.3 shows the observed distribution in time of the occurrence of these extreme events. There is a common feature in many metrics of warm-season extremes in the CONUS - the exceptional heat of the 1920s and especially the 1930s, peaking in 1936. On a per-station average, 60 percent of the Tmax records and 59 percent of the Tmin records occurred in the first half of the period (1899-1961).

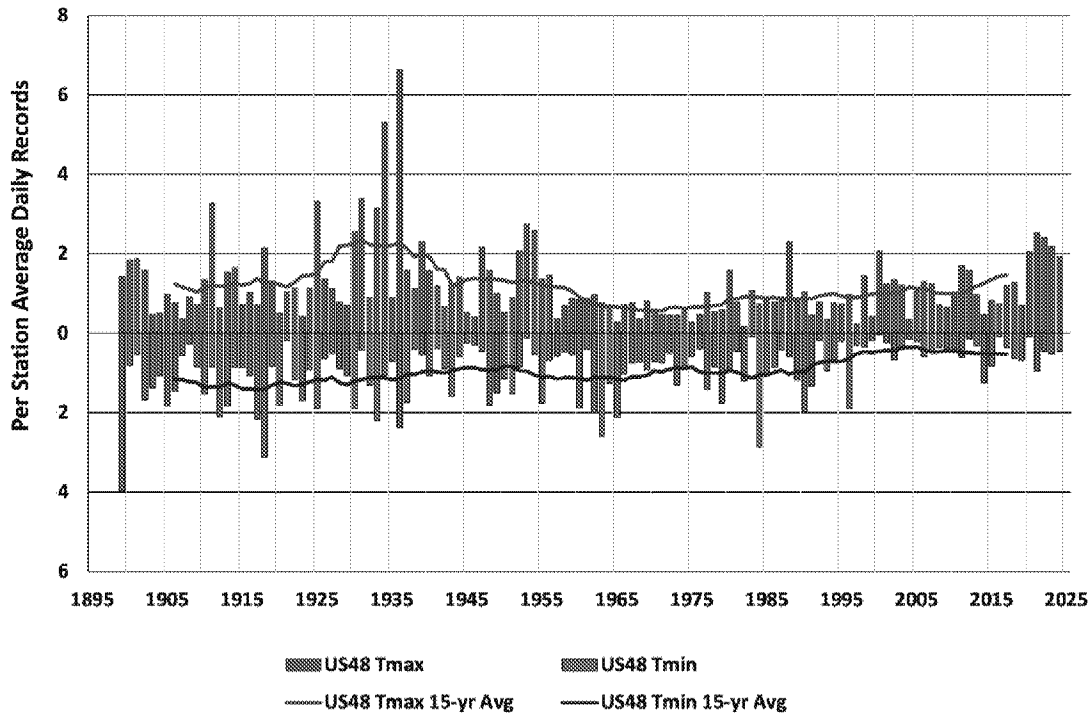


Figure 6.3.3 Number of daily record High (red) and Low (blue) temperatures for warm and cold seasons in the CONUS. The lines represent the 15-year running, centered average. Source: 1,211 USHCN stations supplemented as needed to achieve a minimum of 92 percent of observations in the 126-year period since Dec 1898. US48: contiguous U.S. states. Tmax: maximum temperature. Tmin: minimum temperature.

On the cold side, the Valentine’s Day Arctic outbreak in Feb 1899 stands as the most extensive cold extreme experienced by CONUS, with 1917 in 2nd place. The frequency of cold records has declined, especially over the last quarter of the period in which only 13 percent of the extreme cold events were measured. In contrast, 25 percent of the extreme Tmax records were achieved in the last quarter, in accordance with statistical expectations. These general features have been noted in past assessments (see above, IPCC AR6, NCA4). Combining the two histories, the overall reduction in numbers of both cold and hot extremes over the past century indicates a climate less prone to extremes.

This pattern is also shown in Figure 6.3.4. For each station and for each year the hottest warm season and coldest cold-season temperatures were calculated. Then the differences between these were computed by station and geographically-averaged over all stations, thus yielding an annual measure of the expected range of local extreme temperatures for each year. Figure 6.3.4 shows the 15-year trailing average of this measure, which has clearly declined over the past century.

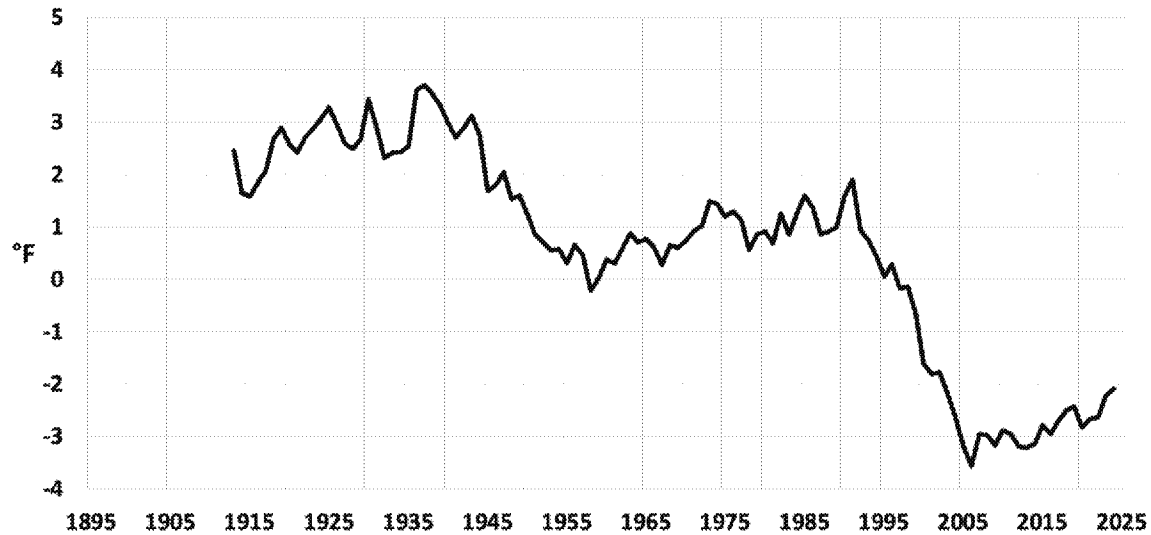


Figure 6.3.4. 15-year trailing average of the difference each year of each station's hottest warm-season Tmax and coldest cold-season Tmin relative to the long-term average. Source: Author analysis of USHCN data.

The average difference for each station between the hottest summer Tmax and coldest winter Tmin has declined by about 5°F in the past 126 years. The decline is due mostly to warmer winter Tmin, but a decline in summer Tmax is also a factor. The rise in Tmin has been strongly related to the growing presence of manufactured surfaces around the weather stations over the last 100+ years (the so-called urban heat island effect; Section 3.3, Karl *et al.* 1988, Runnals and Oke 2006, and Spencer *et al.* 2025).

In summary, while temperature extremes are regularly experienced in the U.S. and attract a great deal of media attention, long term records show the U.S. climate has become less extreme over time (milder) when measured by the range between warm season maxima and cold season minima.

6.3.2 Exceedances of a heat threshold

Under the heading of “The Risk of Temperature Extremes is Changing”, the most recent U.S. National Climate Assessment report (NCA5) notes the increase in a threshold metric of number of days at or above 95°F, stating,

The western U.S. has been particularly affected by extreme heat since the 1980s ..., experiencing a larger increase in days over 95°F, as would be expected given the greater warming in that region relative to the eastern US. Several major heatwaves have affected the U.S. since 2018, including a record-shattering event in the Pacific Northwest in 2021.

Are the occurrences of 95°F days changing? In a climate as varied as that of CONUS, threshold statistics can be misleading. A region with many stations that have near 95°F Tmax average temperatures in the summer might see large swings in the metric when only small changes in average temperature occur. Elsewhere with stations that either rarely or virtually always achieve 95°F Tmax temperatures, a small change will not have much impact on the results.

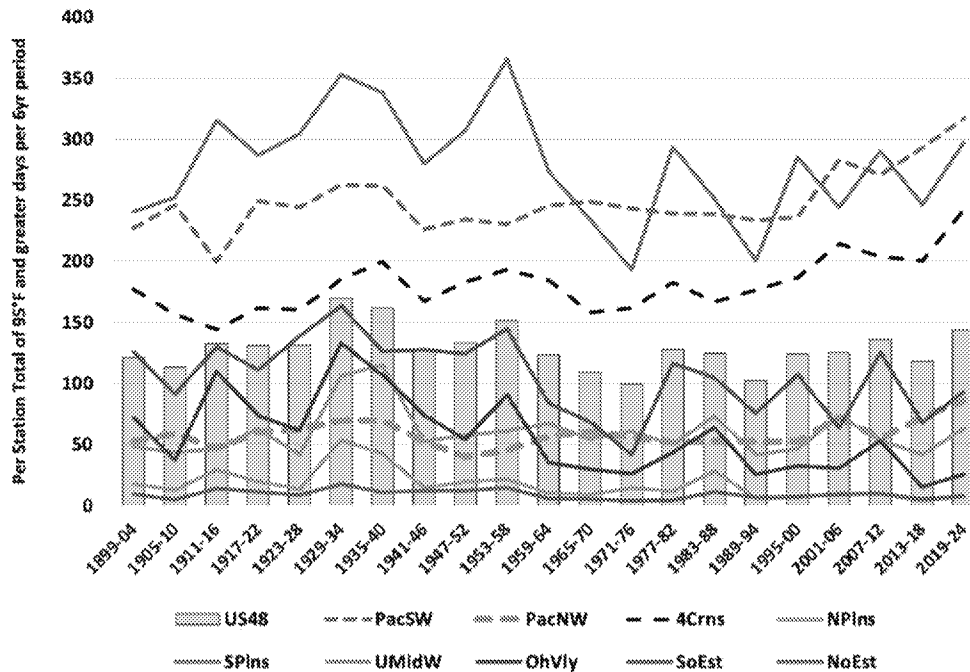


Figure 6.3.5. Total number of days $\geq 95^{\circ}\text{F}$ in 6-yr periods, U.S. 48 (bars) and regions (lines). A 6-year period is used as this evenly divides the 126-year record. Results are robust to using periods from 2 to 11 days. US48: contiguous U.S. states. See Figure 6.3.2 for region names. Source: Author analysis of USHCN data.

In the past 126 years, the average CONUS station experienced 129 days exceeding 95°C per 6-yr period, but the regional values range from 278 in the Southern Plains to 9 in the Northeast. Thus, such threshold analyses must be interpreted with caution. Figure 6.3.5 shows that only three of the nine regions, all in the West, have experienced upward trends in the number of 95°F or hotter days (dashed lines). The CONUS as a whole has not, and the other six regions have experienced declines.

The Pacific NW heatwave of 2021 referenced in the NCA5 quote will be examined more closely in Section 8.6.1. The evidence indicates that it was a single, unprecedented event in the record, not part of a pattern of increasing extreme heat. For example, the 5-day average tropospheric grid-point temperature anomaly over the Pacific NW during that event was $+10.8^{\circ}\text{C}$, the most extreme Northern Hemisphere grid point summer anomaly in the 46 years from over 4 million grid values. In contrast, the global temperature anomaly during that time was virtually zero ($+0.03^{\circ}\text{C}$, Mass *et al.* 2024).

6.3.3 Heatwaves

Heatwaves (consecutive days that exceed an extreme threshold) have a greater societal impact than a single daily record temperature. We measure “Heatwave Days” here as the count of all days in May-Sep each year that exceed the 90th percentile for that day and that lie within a period of at least six consecutive days. This is equivalent to the method used in NCA4, except that the reference period here is the entire record 1899-2024 while NCA4 truncated the reference period to 1961-1990, which was a cool interval. (The pattern of results shown below does not depend on the choice of reference period.) That truncation boosts positive results (days exceeding the 90th percentile) in years warmer than the reference period, especially starting in 1960 and moving to the present (see Figure 6.3.6 and discussion below).

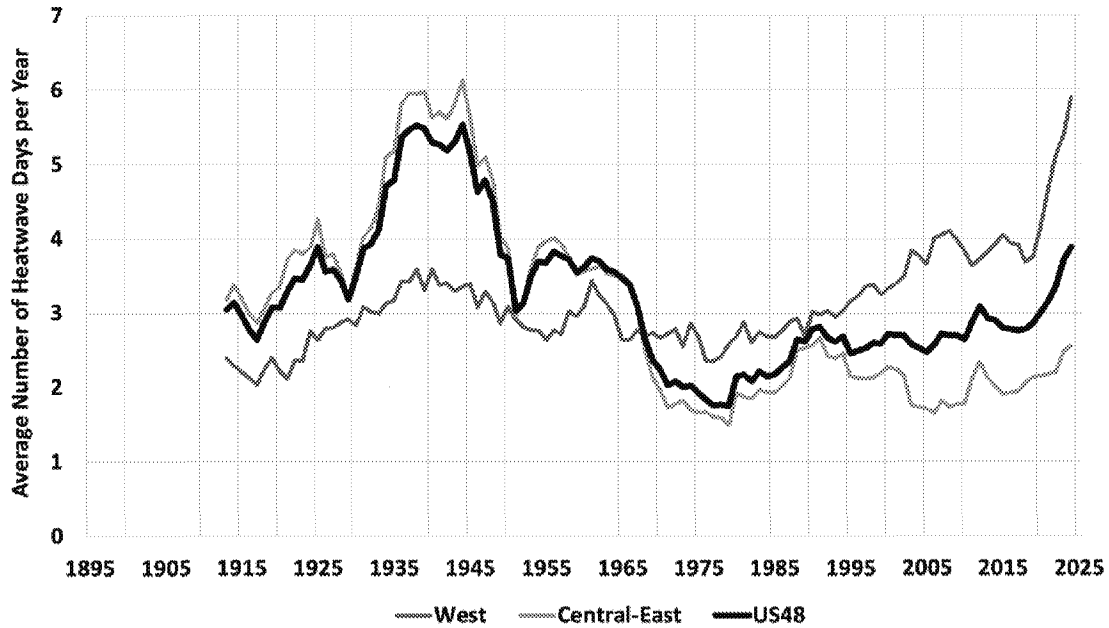


Figure 6.3.6 15-year trailing average of number of heatwave days per year per station in the CONUS (black line) and two regions: West (red), Central-east (green).

Figure 6.3.6 indicates that there are regional variations in heatwave activity. The excessive heat of the first half of the 20th century occurred primarily in the eastern two-thirds of the nation, while the West has seen a recent increase of heatwave days (NCA5). This indicates that the background warm season circulation favored heatwaves in the eastern portions of the country in the first half of the 20th century, but in the 21st century the patterns have favored heatwaves in the West. For CONUS as a whole, heatwaves are no more common today than they were a century ago, consistent with the upper panel of our Figure 6.4.1 taken from NCA4.

This metric varies significantly with region. The four northern regions (Pacific NW, Northern Plains, Upper Midwest, and Northeast) on average experience 15 to 27 heatwave days per 15-year period. In contrast, the five southern regions (Pacific Southwest, 4-Corners, Southern Plains, Ohio Valley and Southeast) see 37 to 54 such days, essentially twice as many. This suggests the summer circulation pattern is more prone to stationary events in the southern regions while transient systems in the northern regions are more common and thus cut short these potentially longer events.

The analysis of heatwaves is an example of why it is important to consider complete datasets and appropriate metrics. The NCA5 directs readers to the website [[HYPERLINK "https://www.globalchange.gov/indicators/heat-waves"](https://www.globalchange.gov/indicators/heat-waves)] (USGCRP 2023) to view a figure showing the number of urban heatwaves by decade from the 1960s, which we reproduce as Figure 6.3.7.

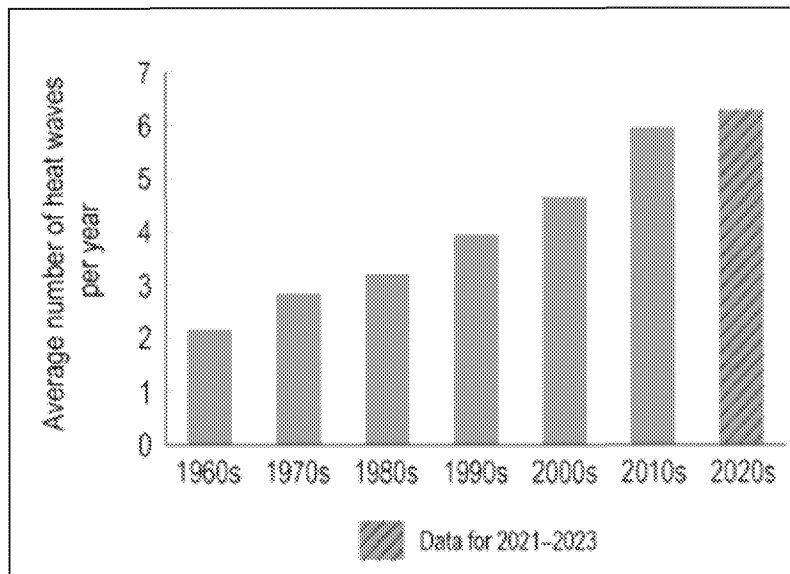


Figure 6.3.7: Average number of urban heatwaves per year for 50 large U.S. metropolitan areas, a misleading metric for reasons explained in the text. From [[HYPERLINK "https://www.globalchange.gov/indicators/heat-waves"](https://www.globalchange.gov/indicators/heat-waves)] (accessed May 22, 2025).

The figure shows a monotonic increase in each decade from two occurrences per year in the 1960s to six in the 2020s. The definition of a heatwave used is an unusual but practical measure of human discomfort - a period of at least 2 consecutive days when the *minimum* apparent temperature (combination of temperature and humidity) exceeds the 85th percentile. Note too, the dataset is limited to the 50 largest U.S. cities.

Given the unusual heatwave definition and urban focus, these increasing values since 1960 presented in USGCRP (2023) are not informative about long term trends or the influence of GHG emissions for at least two reasons. First, as shown in Figures 6.3.1 and 6.3.6, the 1960s was the coldest decade and 1970s the second coldest decade since the 1910's, so this starting date preconditions the time series to show increases. Second, post-1960 urbanization in these cities is a major factor in the rise of T_{min} relative to co-located T_{max} and relative to T_{min} at nearby rural stations (Karl *et al.* 1988, Runnals and Oke 2006, Christy *et al.* 2009, McNider *et al.* 2012). This does not dismiss the real rise in nighttime temperatures in major U. S. cities and the societal impacts associated with these changes. However, we note for a variety of reasons that summer T_{max} (especially in rural areas) is a better metric for detecting changes in heatwaves influenced by changes in the background climate due, for example, to increasing GHGs (Christy *et al.* 2009). For CONUS as a whole, the evidence in this section suggests GHG emissions have had little-to-no effect on heatwaves against the background of urbanization and natural climate variability. Irrespective of the ultimate cause of regional trends, heatwaves have important effects on society that must be addressed, as we discuss in Chapter 10.

BOX: Perils of short data records

San Francisco provides a good case study of the limitation of using short historical samples to characterize natural variability of extreme events. Suppose we use a 130-year sample of daily San Francisco precipitation from 1895 to 2024 and we look for 3-day, 5-day, 14-day and 30-day rainfall records. The results are as shown in Table 6.2.

Event	Record (inches)	Year
3-day	6.94	2023
5-day	8.55	2023
14-day	12.62	2023
30-day	18.93	1998

Table 6.2: Extreme rainfall records, San Francisco, 1895-2024.

The records all cluster in the more recent years. 2023 appears to be an exceptional year and since it is near the end of the sample, it might suggest that the climate has shifted into a more hazardous state, perhaps because of human influences. [Note “2023” indicates the event occurred in the water-year of Aug 2022 to Jul 2023.]

But the picture is very different if we use a sample that begins 45 years earlier, in 1850. Table 6.3 shows that the record-setting events all happened in the 1860s. Furthermore 2023 is now not even in 2nd place but falls to 3rd or 4th place. And comparing the records of the two charts shows the extreme precipitation events in 1862 and 1867 involved considerably more rainfall than the 1998 and 2023 events, with 14- and 30-day totals about 50 percent higher.

Event	Record (inches)	Year	Rank of 130-yr extreme listed in Table 6.2
3-day	8.85	1867	3
5-day	9.80	1867	3
14-day	19.05	1862	4
30-day	28.25	1862	2

Table 6.3: Extreme rainfall records, San Francisco, 1850-2024.

The range of natural variability is made even more remarkable when paleoclimate evidence is examined. Porter *et al.* (2011) discovered that in the past 1,800 years at least six megastorms were more intense than the devastating 1861-62 ARkStorm that struck the region. Evidently such extreme events, “unprecedented” in our 1895-2024 sample, have impacted the region about every 300 years, though not since 1895.

This example illustrates the limitations of using relatively short climate periods (~130 years) to assess the character and range of natural variability in general and of extreme events in particular. Accurate representation of the full range of natural variability is necessary for any attribution analyses (Section 8.6). Infrastructure planners, emergency management institutions and attribution scientists would understand the significant mischaracterization of the magnitude of a future extreme if based only on the last 130 years. In this case, a single time-sample of 130 years provides an underestimation of the extreme value by up to 50 percent determined when adding only 45 more years of observations. Compared to millennial-scale paleoclimate evidence, an even greater underestimation would occur. An important lesson is that the climate can deliver great surprises on its own, even without human influences.

6.4 Extreme precipitation

AR6 assessed that an increase in heavy precipitation has been observed in data starting in the 1950s.

AR6: The frequency and intensity of **heavy precipitation events have increased** since the 1950s over most land area for which observational data are sufficient for trend analysis (high confidence). (SPM A3.2)

AR6: In North America, there is robust evidence that the **magnitude and intensity of extreme precipitation has *very likely* increased** since the 1950s. Both [one-day maxima] and [5-day maxima] have significantly increased in North America during 1950-2018. (Chapter 11, p. 1560)

The U.S. National Climate Assessments (NCA4, NCA5) have highlighted an increase in the occurrence of the heaviest precipitation events (defined in different ways) primarily in the eastern half of CONUS, especially the Northeast, when starting the analysis in either 1901 or 1958. Interestingly, the regional variations indicate that the largest increases in extreme precipitation events are in the Northeast and the smallest in the West, a pattern counter to the changes in temperature extremes (Figure 6.3.6).

McKittrick and Christy (2019) examined long-term and consistent station observations of extreme daily precipitation to test some of these NCA claims for the Southeast and West Coast using a trend model with a non-parametric variance estimator robust to the complex autocorrelation properties of precipitation data. When the time series were extended back in time (as far as 1872 in some cases) or started later (1978), there were no significant trends for either region.

These findings have been updated for this report (McKittrick and Christy 2025) with similarly constructed observations from 29 stations on the CONUS Pacific Coast (1893ff from San Diego CA to Blaine WA) and 24 stations in the humid Southeast (1872ff from Austin TX to Washington DC), also adding 27 stations in the Northeast (1888ff from Buffalo NY to Eastport ME). The locations are shown in Figure 6.4.1. The stations were selected based on availability of long-term high-quality records. The regions are each associated with important features of extreme precipitation behavior: the Pacific coast is associated with landfalling atmospheric rivers for which AR6 cites evidence of increasing activity since 1948 with further increases expected as the world warms (AR6 8.3.2.8.2); the NCA report indicates that the Northeast has experienced the greatest increase in extreme events, and the Southeast is also a place noted in the NCAs as having increased extreme events.

The results of applying the analysis of McKittrick and Christy (2019) were as follows for each region, then followed by further explanation.

Pacific Coast heavy rainfall events

- The average precipitation trend is statistically significant (downwards) in Astoria OR; insignificant elsewhere.
- The trend in rainfall variance is positive and significant in Big Sur CA; insignificant elsewhere.
- The trend in daily maximum precipitation is positive and significant in Aberdeen WA and Big Sur CA and negative and significant in Newport OR (insignificant elsewhere).

- Averaged over all stations in the region, none of these three trend parameters is statistically significant.

The Pacific Coast receives considerable precipitation from Atmospheric River (AR) events which often last more than a day or two (e.g., Gershunov *et al.* 2017, Pan *et al.* 2024). The worst series of such events in recent history was the so-called ARkStorm that occurred during December 1861 and January 1862; it dumped nearly 10 feet of rain in parts of California and submerged the entire Central Valley for weeks under as much as 15 feet of water (Brewer 1930, Null and Hulbert 2007). Additionally, paleoclimate research has found six megastorms more severe than 1861–1862 in California during the last 1800 years, occurring at intervals of 300 years or so (Porter *et al.* 2011).



Figure 6.4.1: Locations of precipitation monitoring stations used in this report. Orange: Pacific coast. Blue: Northeast. Green: Southeast. Data from McKittrick and Christy (2025).

We examine occurrences of 5-day deluges as follows. Taking the Pacific coast as an example, a 130-year span contains 26 5-year intervals. At each location we computed the 5-day precipitation totals throughout the year and selected the 26 highest values across the sample. A single year might have more than one of the 26 heaviest. Each of those can be thought of as 1-in-5yr events. If there are no trends in precipitation, then the total number of these events across all stations should be evenly spread over the years. In Fig. 6.4.2 we show the distribution in time of these events for the Pacific coast. The deluges associated with the massive 1997-98 El Niño event are readily apparent. While erratic, as is typical of such precipitation metrics, there is no indication of a tendency to become more frequent over time.

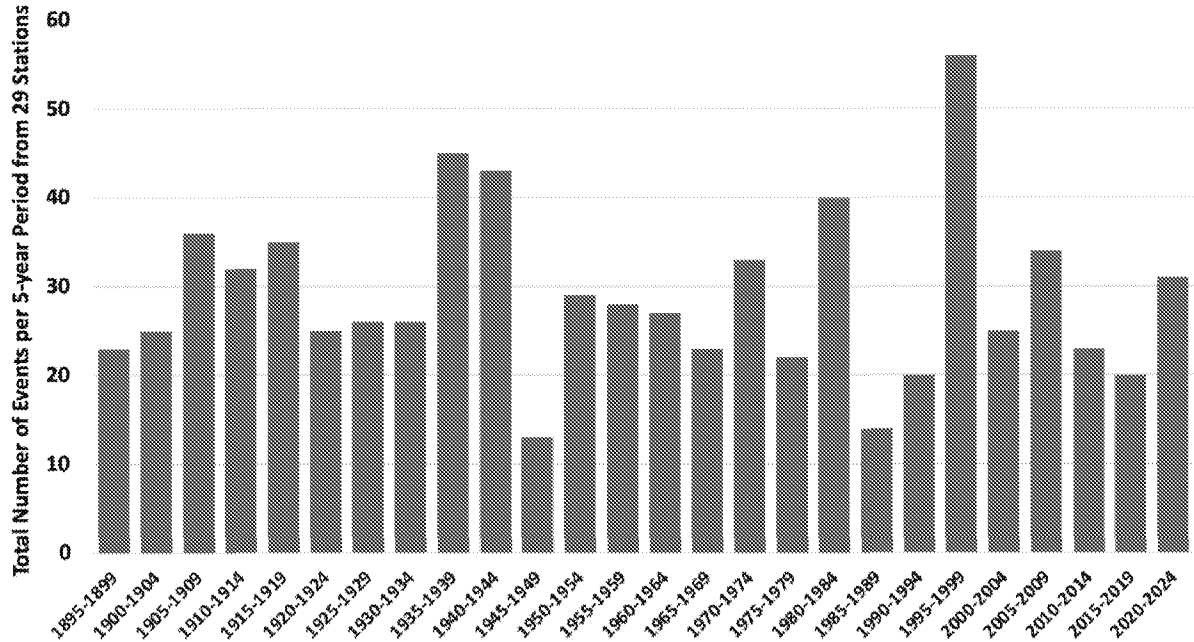


Figure 6.4.2. The time distribution by 5-year periods of the 26 heaviest (1-in-5 yr) occurrences for 29 stations on the Pacific coast.

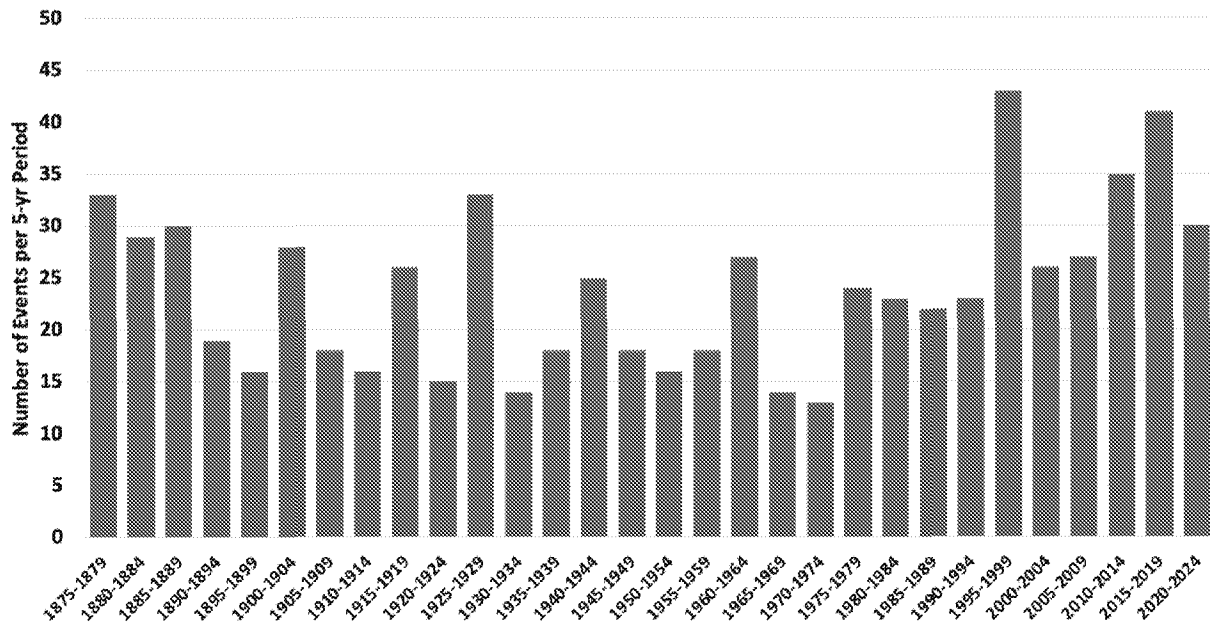


Figure 6.4.3. As in Fig. 6.4.2 but for the heaviest 30 (1-in-5yr) events for 24 stations in the humid Southeast from Austin TX to Washington DC in 5-year bins for 1875-2024.

Southeast heavy rainfall events

- The trend in average precipitation is positive and statistically significant in Mobile AL and Quitman GA but insignificant elsewhere.
- The trend in rainfall variance is positive and significant in Mobile AL but insignificant elsewhere.
- The trend in daily maximum precipitation is positive and significant in Vicksburg MS and Norfolk VA but insignificant elsewhere.
- Averaged over all stations in the region none of these three trend parameters is statistically significant

Figure 6.4.3 is analogous to Figure 6.4.2 for the last 150 years in the Southeast humid zone. The temporal pattern of 5-day totals of the 1-in-5yr heavy events is generally unremarkable, though a cluster of higher values appears in 1995 to 2019. The increase in those years is due largely to the 4 northeastern-most stations of Wilmington NC, Weldon NC, Washington DC, and Norfolk VA. This confirms the pattern indicated in NCA4 and NCA5 -- an increasing frequency of heavy events due to a temporal clustering of tropical storms from eastern NC to Maine discussed below. Otherwise, the remaining 20 stations show an unremarkable temporal distribution of heavy events.

Northeast heavy rainfall events

- The trend in average precipitation is positive and statistically significant in 12 of 27 locations and also in the regional average.
- The trend in rainfall variance is positive and significant in Portland ME, Albany NY, Buffalo NY and Eastport ME but insignificant elsewhere.
- The trend in daily maximum precipitation is positive and significant in Portland ME, Gardiner ME and Eastport ME but insignificant elsewhere.
- When averaged over all stations in the region, there is no statistically significant trend in either the precipitation variance or maximum

Fig. 6.4.4 is analogous to Figure 6.4.2 for the last 135 years in 27 stations in the Northeast (including Montreal Canada). We use 3-day totals here as this produced the largest temporal variations in time. In this region, 77 percent of events occur during June to October and are dominated by incursions of hurricanes, tropical storms, or tropical storms that transition to extratropical systems. According to NCA4 and NCA5 this region experienced the largest increases in extreme events, so it merits a closer examination.

There is a noticeable clustering of extreme events from 1995 to 2014. Howarth *et al.* (2019) examined a similar region as in Fig. 6.4.4 that includes PA and NJ and reported significant differences in various precipitation extremes between two 18-year periods, 1979-96 and 1997-2014. That included a 317 percent increase in 24-hr events exceeding 6 inches, while we find a 58 percent increase over the same years. However, Figure 6.4.4 shows that frequency drops sharply after 2014, returning to the long-term average in the subsequent 5-year intervals, again illustrating the perils of drawing conclusions from short-term trends in highly variable metrics.

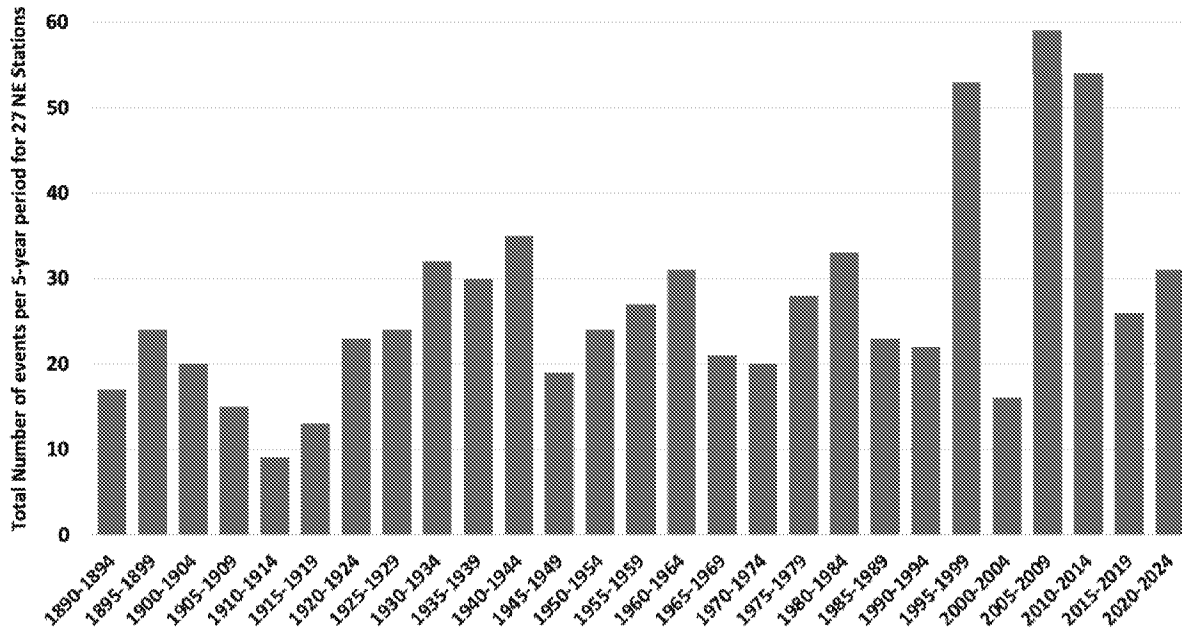


Figure 6.4.4. As in Fig. 6.4.2 but for the heaviest 27 (1-in-5yr) 3-day precipitation events for 27 Northeast stations from NY to ME, including Montreal.

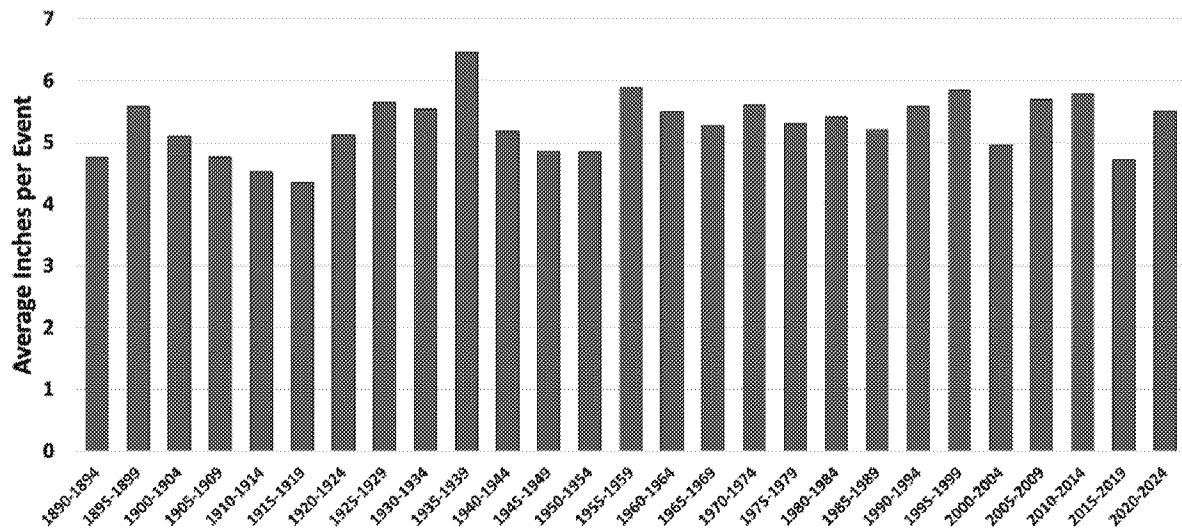


Figure 6.4.5. Average amount of precipitation falling in the 1-in-5yr events for the NE stations.

The high percentage increase in the Howarth *et al.* sample is associated with small numbers at relatively few locations: there were only 6 in the first period and 25 in the second across 58 stations. Most of the stations did not record such an event. If there is a region-wide increase in heavy events, it should be seen in the average across all stations. Figure 6.4.5 shows the average precipitation in the 1-in-5 yr events across the NE. The trend is only +0.04 inches/decade. The highest amount, 1935-39, includes the Great New England Hurricane of 1938, one of the rare major (Category 3 or higher) hurricanes to strike the region.

The results in Figs. 6.4.4 and 6.4.5 thus suggest that though there was a surge in the number of events at a few locations during the 1995 to 2014 interval, there was no regional pattern and the change did not persist beyond 2014.

Jong *et al.* (2024) document the increase of tropical influences on precipitation events in the Northeast since 1959 and concluded “The autumn extreme precipitation trend over the Northeast U.S. is primarily attributed to tropical cyclone-related events since the 1990s.” The question then becomes: “Was the temporal clustering of tropical systems in 1997-2014 which affected the Northeast a response to increasing GHGs?” Jong *et al.* examined CMIP-6 model output which suggests that there will be fewer such systems in the 21st century but that the intensity of the rainfall events might increase. This conjecture is not seen in Fig. 6.4.5 where the amount-per-event has remained steady over the 135-year period.

There is some evidence to indicate the heaviest rainfall events might be redistributed due to the impact of urban infrastructure on the local weather (e.g., Pielke Sr. *et al.* 2011, Zhang *et al.* 2018, Yang *et al.* 2024). Yang *et al.* state “Cities that experience compact development tend to witness larger increases in extreme rainfall frequency over downtown than their rural surroundings, while the anomalies in extreme rainfall frequency diminish for cities with dispersed development.” While this is an important insight to consider, the effect on the specific stations used in this analysis is unknown, or at least not detectable in Fig. 6.4.5.

In summary, some U.S. regions show short-duration increases in extreme precipitation events, consistent with natural variability. But analysis of long term, nationwide historical records that considers the autocorrelation properties of precipitation data does not support the claim that extreme short-duration rainfall events are becoming more frequent or intense.

6.5 Tornadoes

AR6 assesses tornado trends in the U.S. as follows:

[O]bservational trends in tornadoes, hail, and lightning associated with severe convective storms are not robustly detected due to insufficient coverage of the long-term observations. There is *medium confidence* that the mean annual number of tornadoes in the USA has remained relatively constant. (Chapter 11, section 11.7.3, p. 1594)

The monitoring of weak tornadoes has changed over time. The growth of rural populations and the increasing ability to take video with hand-held devices has led to more frequent reports of weak tornadoes that produce minimal damage. In contrast, strong to violent tornadoes have been observed more consistently over time. Note that tornado strength is measured by the damage it produces, not by the visual appearance of the funnel. Limited real-time observational capabilities in earlier decades did not prevent identification because strong to violent tornadoes leave much more damage which will be assessed later even if the tornado itself was not observed. Since statistics began in 1950, there has been a substantial decrease (by about 50%) in the number of strong to violent tornadoes as shown in Fig. 6.5.1a.

To summarize, there is a noticeable downward trend in the number of severe tornadoes in the U.S. since 1950. After 1990 the number of weak tornadoes in the U.S. has remained roughly constant; data before that are incomplete due to limited monitoring.

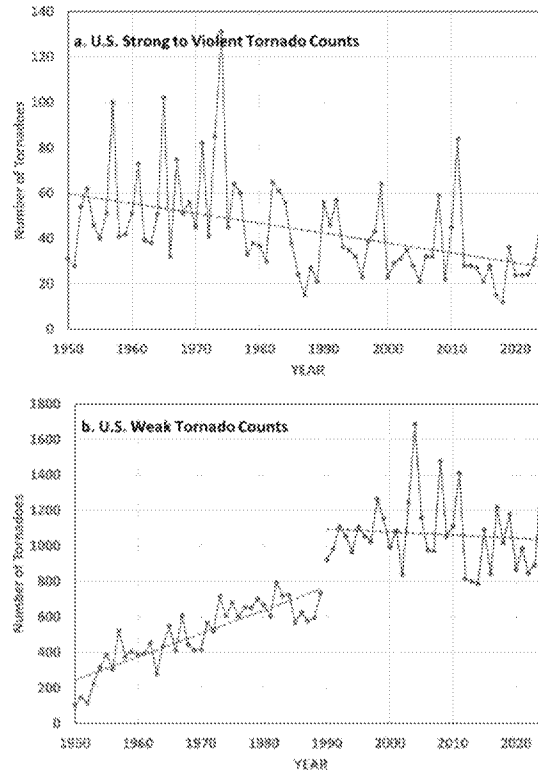


Figure 6.5.1. Annual U.S. tornado counts for (a) strong to violent tornadoes (EF3 to EF5), and (b) weak tornadoes (EF0 to EF2). Based upon NOAA Storms Prediction Center data, available at https://www.spc.noaa.gov/wcm/data/1950-2024_actual_tornadoes.csv

6.6 Flooding

Changes in floods were assessed as follows:

AR6: The SREX **assessed low confidence for observed changes in the magnitude or frequency of floods** at the global scale. This assessment was confirmed by the AR5 report. The SR15 found increases in flood frequency and extreme streamflow in some regions, but decreases in other regions. . . [H]ydrological literature on observed flood changes is heterogeneous, focusing at regional and sub-regional basin scales, making it difficult to synthesise at the global and sometimes regional scales. (Chapter 11.5)

AR6: [T]he seasonality of floods has changed in cold regions where snowmelt dominates the flow regime in response to warming (*high confidence*). Confidence about peak flow trends over past decades on the global scale is *low*. (Chapter 11.5)

NCA4: Trends in extreme high values of streamflow are **mixed** across the United States. Analysis of 200 U.S. stream gauges indicates areas of both increasing and decreasing flooding magnitude **but does not provide robust evidence** that these trends are attributable to human influences (pp. 240-241)

The absence of detectable US-wide trends in flooding is consistent with the findings in Section 6.4 of absence of coherent changes in extreme precipitation.

6.7 Droughts

Assessments of drought trends were as follows.

AR6: Few AR6 regions show observed increases in meteorological drought (Section 11.9, p. 1575),

AR6: Increasing trends in agricultural and ecological droughts have been observed on all continents (*medium confidence*), but decreases only in one AR6 region (*medium confidence*). Increasing trends in hydrological droughts have been observed in a few AR6 regions. (Chapter 11 Summary)

NCA4: As a consequence of this increased precipitation, **drought statistics over the entire CONUS have declined.** (p. 233)

NCA4: Recent droughts and associated heat waves have reached record intensity in some regions of the United States; however, by geographical scale and duration, the Dust Bowl era of the **1930s remains the benchmark** drought and extreme heat event in the historical record (very high confidence). (p.231)

SREX: From a paleoclimate perspective, **recent droughts are not unprecedented**, with severe ‘megadroughts’ reported in the paleoclimatic record for Europe, North America, and Australia. (p. 170)

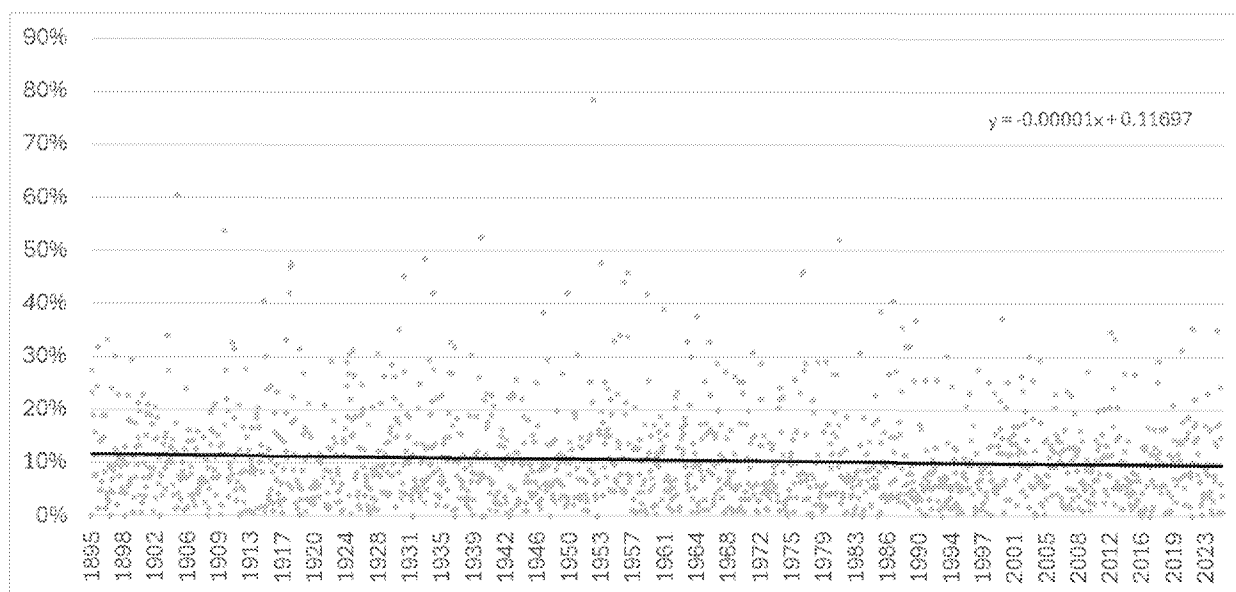


Figure 6.7.1: Monthly percent of US classified as “Very Dry” 1895—2025. Data source: NOAA [[HYPERLINK "https://www.ncei.noaa.gov/access/monitoring/uspa/wet-dry/0"](https://www.ncei.noaa.gov/access/monitoring/uspa/wet-dry/0)] least squares trendline added.

As shown in Figure 6.7.1, U.S. long-term data shows an insignificantly declining trend in extreme dryness (-0.001 percent per year)

Kogan *et al.* (2020) examines a 38-year high-resolution satellite-based drought measure and concludes that global drought has not intensified and is not connected to climate change: “it is possible to state firmly that global and main grain countries’ drought area and intensity trends have not been following global climate warming since 1980’s.”

In summary there is no evidence of increasing meteorological drought frequency or intensity in the U.S. or globally over recent decades.

6.8 Wildfires

The IPCC has not provided an attribution assessment of wildfires. As shown in Figure 6.8.1, global wildfire activity as measured by European Space Agency shows a downward trend in the 21st century.

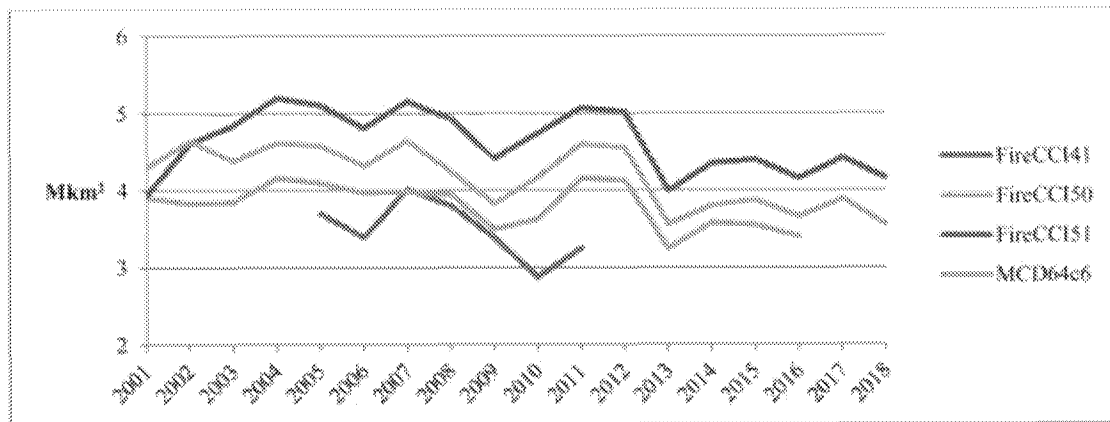


Figure 6.8.1: Global wildfire area 2001-2018. Source: From Lizundia-Loiola *et al.* (2021) Figure 12. The different coloured lines represent data products derived from different satellites and algorithms

Global data show that wildfire coverage is constant or declining on every continent (Samborska and Ritchie, 2024). However there is evidence that the intensity of fires in some regions is worsening (Cunningham *et al.* 2024) and that wildfires resulted in a net loss of global forest cover over 2001-2019 (Tyukavina *et al.* 2022).

Active fire suppression since 1900 makes it difficult to establish a natural baseline for wildfire activity in the U.S. Paleoclimatic evidence indicates that past activity was much higher than today. Marlon *et al.* (2012) used sedimentary charcoal layers to reconstruct fire history of western U.S. for the past 1400 years and also fit a model to predict fire activity as a function of climatic conditions. Their results are summarized in Figure 6.8.2 below (from Figure 2 in their paper). There has been a growing wildfire deficit over the 20th century. In other words, however much fire was observed in the 20th century, it was less than what would have been observed in previous centuries based on the climatic conditions. Parks *et al.* (2025) likewise find that despite the recent increase in wildfire burn area in North America, a significant wildfire deficit remains relative to historical wildfire regimes.

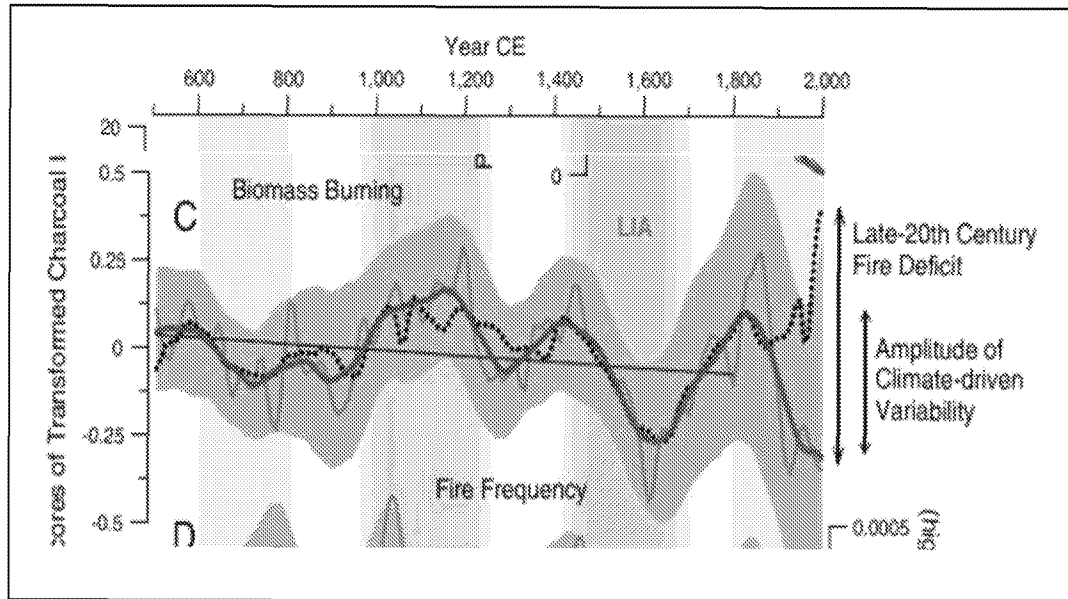


Figure 6.8.2 Fire frequency and fire deficit in the US. The red line shows the smoothed charcoal record and the black dotted line shows the predicted charcoal record based on climatic conditions. Source: Marlon *et al.* (2012) Figure 2.

U.S. data from the National Interagency Fire Centre (NIFC) from 1926 to 2023 are shown in Figure 6.8.3. The NIFC has removed the pre-1960 data from its current website on the grounds that measurement methods changed after 1960 making the comparison unreliable. Nonetheless just focusing on the post-1985 interval the number of fires is not increasing. The area burned did increase but only until about 2007.

Forest fires have always been a part of nature, and they can certainly create conditions that are inhospitable in the short term for all life, including humans. Science has confirmed the overall benefit and necessity of forest fires. While recent high-profile fires and seasons serve as a reminder of the potential destructive impact, the highest profile U.S. forest fire remains the 1910 Big Blowup fire in the U.S. west, which destroyed over three million acres and eliminated entire towns like Taft, MT (Apple, 2020). The 1910 fire reshaped the U.S. Forest Service (National Forest Foundation 2022) leading to a focus on fire suppression with a primary goal of defeating all forest fires (Forest History Society, 2022). This led to the “10 am rule” in 1935 requiring that all fires spotted on any day had to be controlled by 10 am the following day (National Forest Foundation, 2022).

While defeating all fires seemed a noble goal, questions began to arise as to whether this behavior “followed the science” (U.S. Forest Service, 2022). Over time the U.S. Forest Service has begun to rethink its goals, recognizing that new approaches such as prescribed burns, fuel elimination, and controlled wildfires are more appropriate (Sommer, 2016). Recent research is validating this approach and recognizing that more frequent smaller fires likely result in healthier forests, water ecosystems and biodiversity (Stephens *et al.*, 2021).

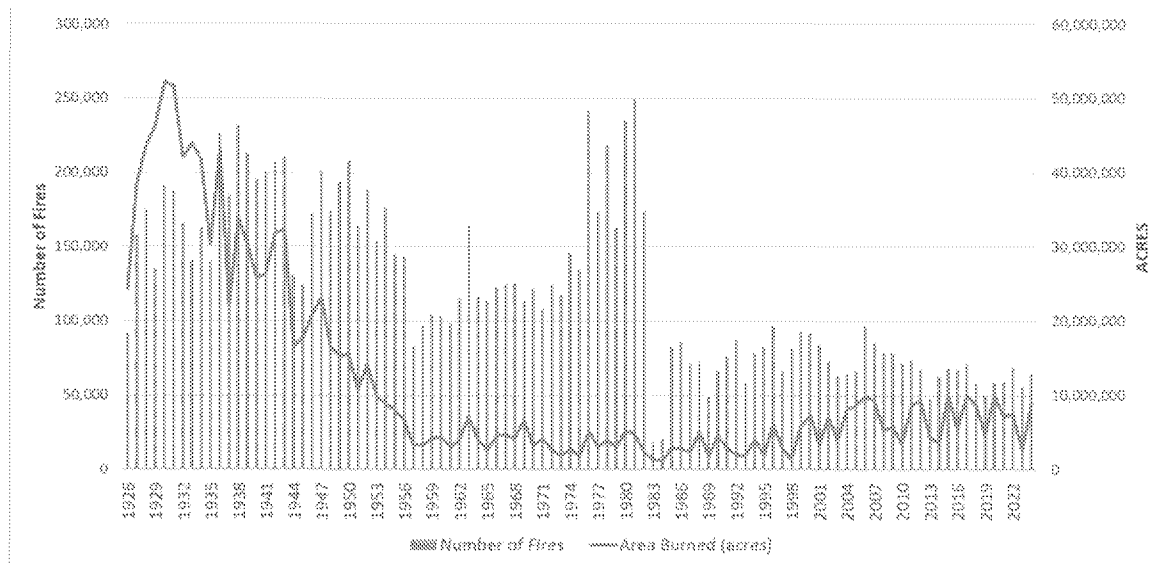


Figure 6.8.3: U.S. wildfires 1926 to 2023. Source: Post-2018: National InterAgency Fire Center data [[HYPERLINK "https://www.nifc.gov/fire-information/statistics/wildfires"](https://www.nifc.gov/fire-information/statistics/wildfires)]. Pre-2017 webarchive.org (n.d.).

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7 CHANGES IN SEA LEVEL

Chapter Summary

Since 1900, global average sea level has risen by about 8 inches. Sea level change along U.S. coasts is highly variable, associated with local variations in processes that contribute to sinking and also with ocean circulation patterns. The largest sea level increases along U.S. coasts are Galveston, New Orleans, and the Chesapeake Bay regions - each of these locations is associated with substantial local land sinking (subsidence) unrelated to climate change.

Extreme projections of global sea level rise are associated with an implausible extreme emissions scenario and inclusion of poorly understood processes associated with hypothetical ice sheet instabilities. In evaluating AR6 projections to 2050 (with reference to the baseline period 1995-2014), almost half of the interval has elapsed by 2025, with sea level rising at a lower rate than predicted. U.S. tide gauge measurements reveal no obvious acceleration beyond the historical average rate of sea level rise.

7.1 Global sea level rise

Global sea level rise is arguably the most important climate impact driver that is unambiguously associated with increasing temperatures. At the global level, warming raises sea level through thermal expansion of sea water and through melting of glaciers and ice sheets. Variations in land water storage are another important factor. At the regional scale, sea level change is influenced by large-scale ocean circulation patterns, and geologic processes and deformation from the redistribution of ice and water. Locally, vertical land motion from geologic processes, ground water withdrawal, and fossil fuel extraction are also important.

AR6 estimates that global mean sea level increased by 7.9 (5.9–9.8) inches between 1901 and 2018, with the rate of sea level rise accelerating in recent decades. At the ocean basin scale, sea levels have risen fastest in the Western Pacific and slowest in the Eastern Pacific over the period 1993–2018 (Fox-Kemper *et al.*, 2021). The rate of global sea level rise is estimated to be 0.12 inches/year, about the height of two stacked pennies (NASA, 2020).

The observing systems for global sea level rise have advanced significantly in the satellite era, particularly with the advent of satellite altimeters in 1993. Local tide gauges have provided useful data for the past century, and even longer for a few locations. Following the end of the Little Ice Age in the mid-nineteenth century, tide gauges show that the global mean sea level began rising during the period 1820–1860, well before most anthropogenic greenhouse gas emissions.

7.2 U.S. sea level rise

Observed and predicted rates of mean global sea level rise might have little scientific relevance for specific locations, owing to local processes (NOAA, 2025). Figure 7.1 shows that in Canada and Alaska (and also northern Washington), sea level is decreasing, owing to uplift from glacial rebound. Most of the Pacific coast tide gauges show low rates of sea level rise, while largest U.S. rates are on the Gulf coast (Louisiana and Texas) and in the mid-Atlantic states (Chesapeake Bay region).

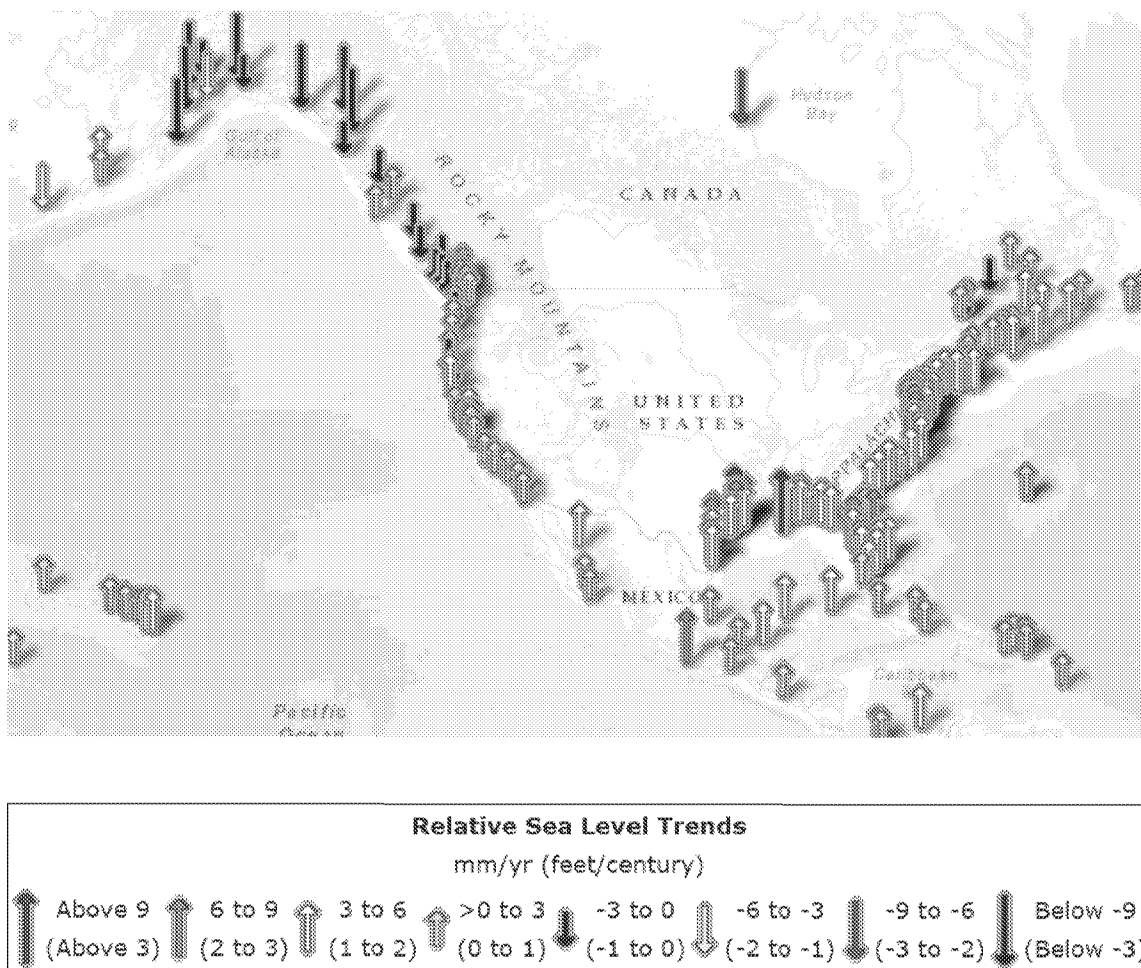


Figure 7.1. Map of rates of relative sea level rise along the U.S. coast (NOAA, <https://tidesandcurrents.noaa.gov/sltrends/>). For reference, 3 mm = 0.12 in.

Measurements of relative sea level rise from tide gauges conflate the climatologically relevant increase in the volume of seawater with local vertical land motion. The latter, which varies from place to place, is best measured using a Global Positioning System (GPS) station located near the tide gauge. It is driven by a range of processes that can be comparable to the climate signal and can locally exacerbate (subsidence/sinking) or mitigate (uplift) the risk of sea level rise (Wöppelmann and Marcos, 2016). Human activities triggering subsidence are often significant. They include soil drainage (*e.g.* for urban development) and subsurface extraction of groundwater or hydrocarbons.

Table 7.1 shows absolute sea-level rise (ASLR) for selected locations, determined from the sum of uncorrected relative sea-level rise (RSLR) as estimated from tide gauge time series (NOAA, 2025) and the vertical land motion (VLM) measurements (NAS, 2012; Letetrel *et al.*, 2015; Karegar *et al.*, 2016). The absolute sea level rise for each of these locations is significantly smaller than the measured relative sea level rise owing to local subsidence. More than half of the measured relative sea level rise is attributed to land sinking for these locations: San Francisco, Galveston, Grand Isle. For reference, the global average rate of absolute sea level rise is estimated to be 0.12 inches/year.

Location	RSLR	VLM	ASLR
San Francisco, CA	+0.08	-0.06	+0.02
Galveston, TX	+0.26	-0.19	+0.07
Grand Isle, LA	+0.36	-0.28	+0.08
St Petersburg, FL	+0.12	-0.02	+0.10
New York City, NY	+0.11	-0.05	+0.06

Table 7.1 Absolute sea level rise (inches/year) consisting of Relative Sea Level Rise (RSLR) plus Vertical Land Motion (VLM)

San Francisco Bay

Over the past 100 years, relative sea level in the San Francisco Bay area has risen by 7.8 inches, at an average rate of 0.08 inches/year (Figure 7.2). As shown in Table 7.1, San Francisco’s vertical land motion is -0.06 inches/year (sinking), producing a recent absolute rate of +0.02 inches/year. Portions of Treasure Island, San Francisco International Airport, and Foster City are sinking as fast as 0.4 inches/year (Shirzaei and Bürgmann, 2018). Problems in the San Francisco Bay area, including threats to the airport, are caused primarily by soil compaction in landfill zones that were formerly wetlands, not by the slow creep of global sea level rise.

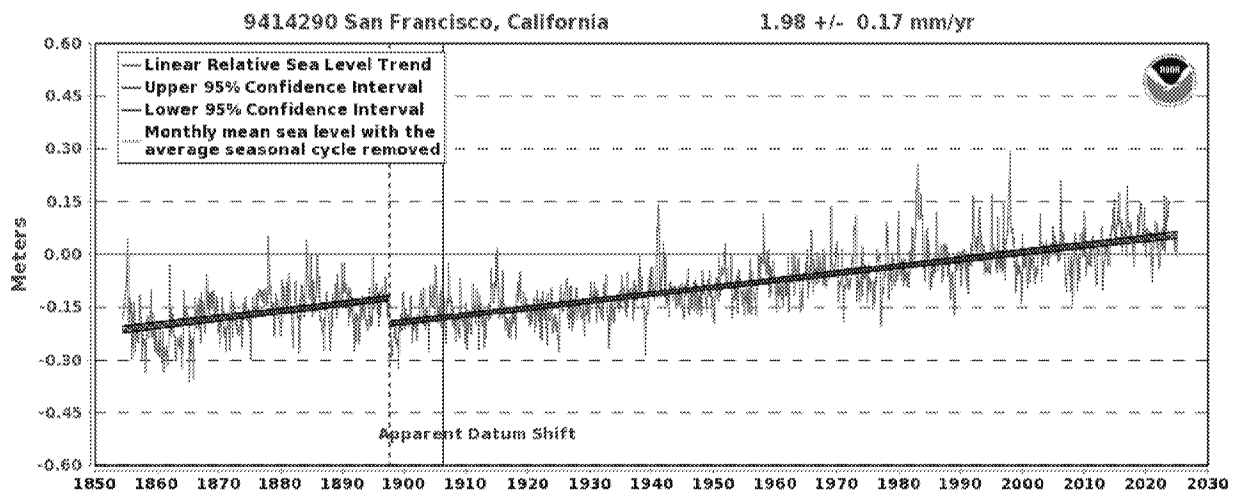


Figure 7.2. Tide gauge measurements at San Francisco, California, obtained from NOAA - https://tidesandcurrents.noaa.gov/sltrends/sltrends_station.shtml?id=9414290 (downloaded 4/22/25).

Galveston - Houston

Long-term tide measurements at Galveston Pier 21 show sea level rise of 2.18 feet in the past century, or a rate of 0.26 inches/year (Figure 7.3). Vertical land motion (subsidence) at Galveston is estimated at -0.19 inches/year, yielding an absolute rate of rise of +0.07 inches/year (Table 7.1). The U.S. Geologic Survey found that most of the land-surface subsidence in the Houston-Galveston region is a direct result of groundwater withdrawals (Kasmarek and Ramage 2017), which caused compaction of the aquifer sediments, mostly in the fine-grained silt and clay layers. By 1979, as much as 10 feet of subsidence had occurred in Houston.

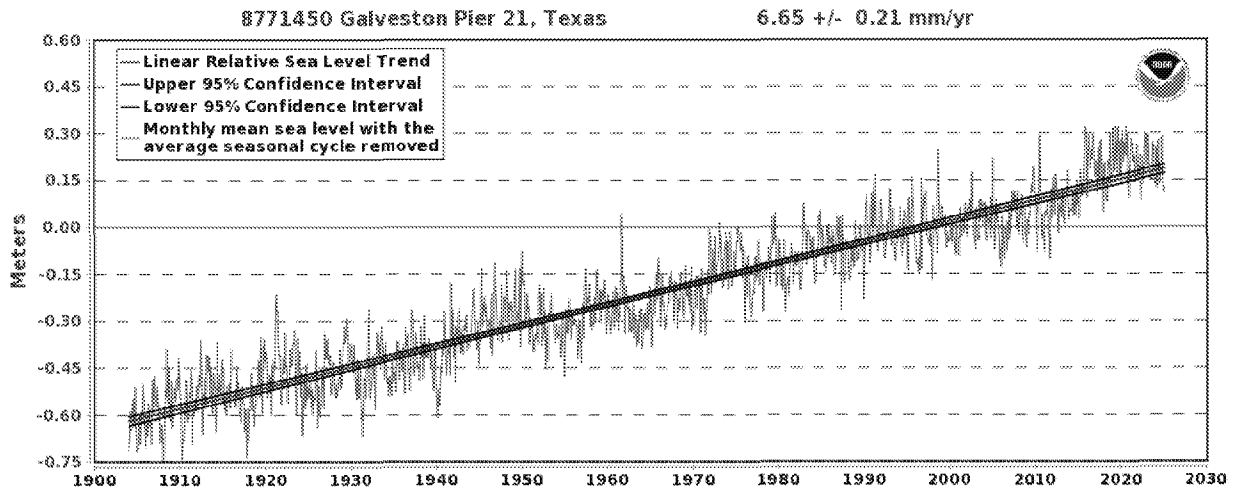


Figure 7.3. Tide gauge measurements Galveston Pier, TX, obtained from NOAA - https://tidesandcurrents.noaa.gov/sltrends/sltrends_station.shtml?id=8771450 (downloaded 4/22/2025).

New Orleans and the Mississippi delta

Long-term tide gauge measurements at Grand Isle, Louisiana, show that sea level has risen by slightly more than 3 feet over the last 100 years at an average rate of 0.36 inches/year (Figure 7.4). Vertical land motion (subsidence) is estimated -0.28 inches/year. Table 7.1 gives the absolute sea level rise as +0.08 inches/year.

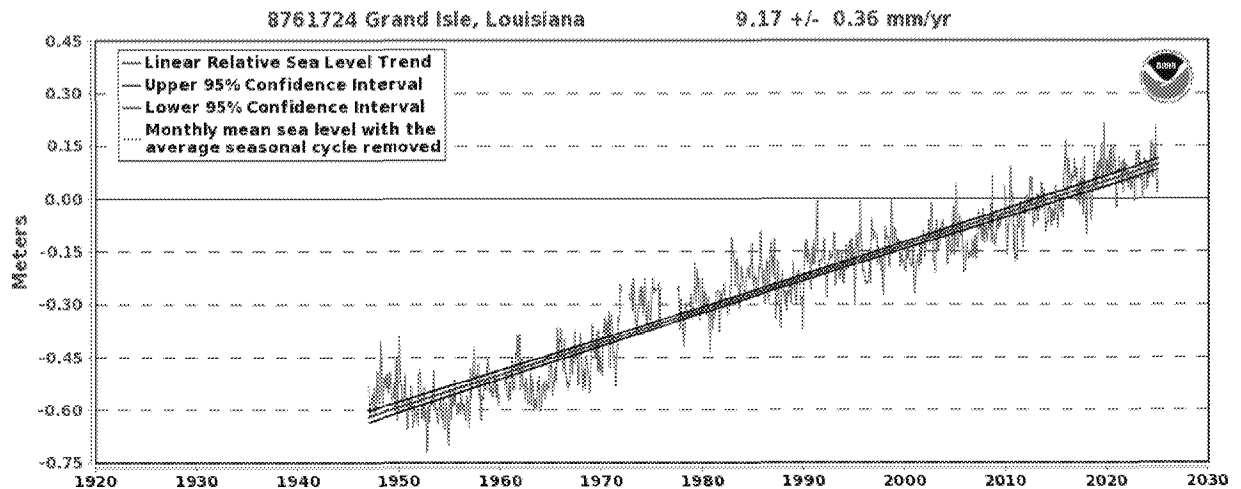


Figure 7.4. Tide gauge measurements at Grand Isle, LA, obtained from NOAA - https://tidesandcurrents.noaa.gov/sltrends/sltrends_station.shtml?id=8761724 (downloaded 4/22/25).

The issues of sea level rise and land loss in the Mississippi Delta region are complex, with geological subsidence and the decline in sediment transported by the Mississippi River being the dominant drivers. The construction of dams in the basin since the 1950s has decreased the Mississippi’s suspended sediment load by ~50 percent (Maloney 2018). A new subsidence map of coastal Louisiana finds the coastal region to be sinking at about one third of an inch per year (Neinhuis *et al.* 2017), associated with groundwater and resource withdrawal. As the city’s elevation averages one to two feet *below* sea level, sea level rise from anthropogenic warming is hardly the dominant driver of New Orleans’ problems.

New York City

New York City is particularly vulnerable to the effects of sea level rise because it is built primarily on islands and has 520 miles of coastline. Measurements by a tide gauge at the southern tip of Manhattan (The Battery) show that relative sea level has risen over 11 inches over the past century, at an average rate of 0.11 inches/year (Figure 7.5). But vertical land motion in the New York City area is -0.05 inches/year (roughly 5 inches per century), so that the absolute rate of sea level rise at The Battery is 0.06 inches/year, or about 55 percent of the measured relative sea level rise.

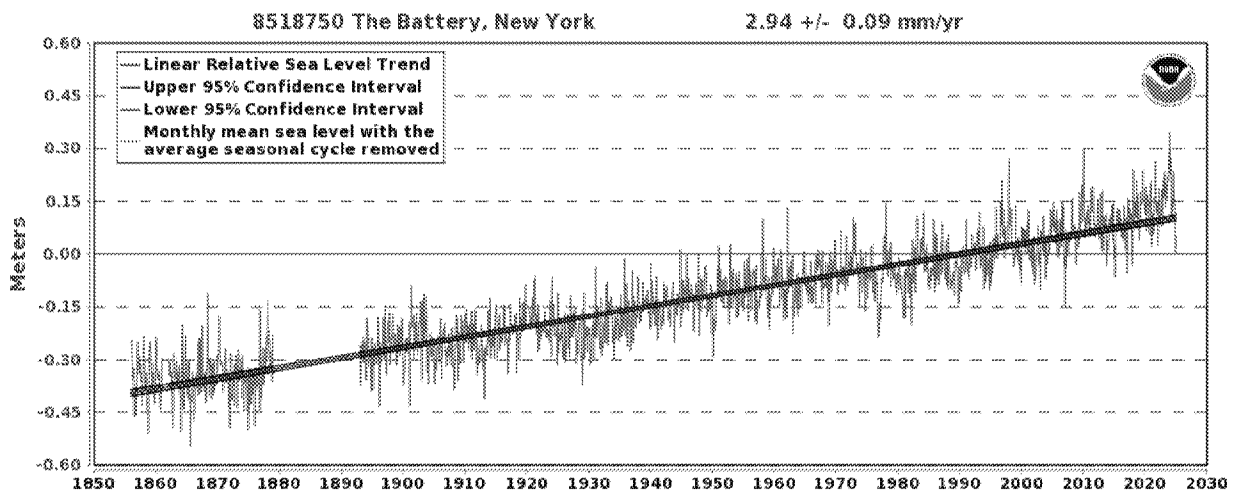


Figure 7.5. Tide gauge measurements at The Battery, New York, obtained from NOAA - https://tidesandcurrents.noaa.gov/sltrends/sltrends_station.shtml?id=8518750 (downloaded 4/22/25).

7.3 Projected sea level rise

The concern over sea level rise is not about the roughly eight inches of global rise since 1900. Rather, it is about projections of *accelerated* rise based upon simulations of a warming climate through the 21st century.

AR6 finds *high agreement* across published global mean sea level projections for 2050 with little sensitivity to emissions scenarios. Considering only projections incorporating ice-sheet processes in whose quantification there is at least *medium confidence*, the global sea level projections for 2050, across all emissions scenarios, fall between 3.94 and 15.75 inches (5th–95th percentile *very likely* range) relative to the 1995–2014 baseline period (Fox-Kemper *et al.*, 2021).

Conversely, AR6 states there is *low agreement* across published global mean sea level projections for 2100, particularly for higher emissions scenarios. Considering only projections representing ice sheet processes in whose quantification there is at least *medium confidence*, the AR6 global mean sea level projections for 2100 lie between 7.9 and 39.4 inches (5th–95th percentile *very likely* range) under the medium emissions scenario SSP2–4.5 (Fox-Kemper *et al.*, 2021). There is deep uncertainty surrounding projections of sea level rise to 2100 owing to uncertainties in ice sheet instabilities, particularly for the higher emissions scenarios.

In February 2022, NOAA issued its projections of sea level rise for various sites along the U.S. coast (Sweet *et al.*, 2022). They claim that by 2050, the sea will have risen one foot at The Battery in Manhattan (relative to 2020). A one-foot rise in thirty years would be more than twice the current rate and about three times the average rate over the past century. In that historical context, NOAA’s projection is remarkable—as shown in Figure 7.6, it would require a dramatic acceleration beyond anything observed since the early 20th century. But even more noteworthy is that Sweet *et al.* (2022) say this rise is “locked in”—it will happen no matter what future emissions are. We should know in a decade or so whether that prediction has legs.

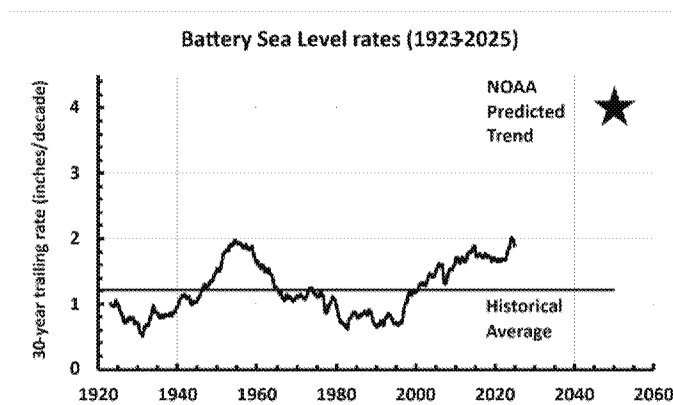


Figure 7.6 Rate of sea level rise at the Battery in Manhattan. Shown is the historical thirty-year trailing trend, together with the allegedly “locked in” NOAA predicted trend for 2050. Historical data: NOAA Tides and Current.

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8 UNCERTAINTIES IN CLIMATE CHANGE ATTRIBUTION

Chapter summary

“Attribution” refers to identifying the cause of some aspect of climate change, specifically with reference to anthropogenic activity. There is an ongoing scientific debate around attribution methods, particularly regarding extreme weather events. Attribution is made difficult by high natural variability, the relatively small expected anthropogenic signal, lack of high-quality data, and reliance on deficient climate models. The IPCC has long cautioned that methods to establish causality in climate science are inherently uncertain and ultimately depend on expert judgement.

Substantive criticism of the main IPCC assessments of the role of CO₂ in recent warming focus on inadequate assessment of natural climate variability, uncertainties in measurement of solar variability and in aerosol forcing, and problems in the statistical methods used for attribution.

The IPCC does not make attribution claims for most climate impact drivers related to extreme events. Statements related to statistics of global extremes (*e.g.* event probability or return times, magnitude and frequency) are not generally considered accurate owing to data limitations and are made with low confidence. Attribution of individual extreme weather events is challenging due to their rarity. Conflicting claims about the causes of the 2021 Western North America Heatwave illustrate the perils of hasty attribution claims about individual extreme events.

8.1 Introduction

The Intergovernmental Panel on Climate Change (IPCC) distinguishes between detection of climate change and attribution of its cause. As defined by the AR6 WGI Glossary (IPCC, 2025):

Detection: Detection of change is defined as the process of demonstrating that climate or a system affected by climate has changed in some defined statistical sense, without providing a reason for that change. An identified change is detected in observations if its likelihood of occurrence by chance due to internal variability alone is determined to be small, for example, <10%.

Attribution: Attribution is defined as the process of evaluating the relative contributions of multiple causal factors to a change or event with a formal assessment of confidence.

Both detection and attribution rely on statistical analysis. Detection focuses on whether changes are significant enough to stand out from random variation regardless of cause. Attribution involves comparison of observed events to model-generated counterfactuals. Since experimentation on the climate is not possible, attribution requires statistical inference to assess how much human activities (such as GHG emissions) versus natural factors (like volcanic eruptions) contribute to observed changes. Attribution methods assume all external and internal drivers of the system are known and represented.

There are ongoing scientific debates around attribution methods, especially those for attributing extreme weather events to climate change. The IPCC has long cautioned that methods to establish causality in climate science are inherently uncertain and ultimately depend on expert judgement. AR4 Working Group I (Hegerl *et al.*, 2007) explained it as follows:

‘Attribution’ of causes of climate change is the process of establishing the most likely causes for the detected change with some defined level of confidence... unequivocal attribution would require controlled experimentation with the climate system. Since that is not possible, in practice attribution of anthropogenic climate change is understood to mean demonstration that a detected change is ‘consistent with the estimated responses to the given combination of anthropogenic and natural forcing’ and ‘not consistent with alternative, physically plausible explanations of recent climate change that exclude important elements of the given combination of forcings’ (IPCC, 2001)... The approaches used in detection and attribution research described above cannot fully account for all uncertainties, and thus ultimately expert judgement is required to give a calibrated assessment of whether a specific cause is responsible for a given climate change.

AR5 Working Group II (Cramer *et al.*, 2014) makes the following statement:

Two broad challenges to the detection and attribution of climate change impacts relate to observations and process understanding. On the observational side, high-quality, long-term data relating to natural and human systems and the multiple factors affecting them are rare. In addition, the detection and attribution of climate change impacts requires an understanding of the processes by which climate change, in conjunction with other factors, may affect the system in question.

Because of the complexity of the causal chains in the climate system, investigation of these relationships is exceptionally challenging.

8.2 Attribution methods

The IPCC employs several attribution methods to assess the causes of observed climate changes, distinguishing between natural and human-induced factors. Below is a concise description of the key IPCC attribution methods.

Optimal Fingerprinting uses linear regression to explain variations in observed climate data as a weighted sum of climate model simulations run with and without anthropogenic forcings. The data used in the regression model is weighted to try to minimize the influence of random noise and the estimation method is chosen to account for climate model error.

Time Series Analysis exploits statistical differences between anthropogenic forcing and natural variability to see which dominates observed temperatures and also uses variations in the timing of changes to determine if causal dependence across data series can be inferred.

Process-Based Attribution focuses on understanding the physical mechanisms behind observed changes. This approach combines observations, climate models, and theoretical understanding to attribute changes in specific processes to forcings. This approach is often used for regional climate phenomena, such as monsoon changes or polar amplification

Extreme Event Attribution assesses the role of human influence in the likelihood of occurrence or intensity of extreme weather events (*e.g.* heat waves or droughts). Methods include:

- Probabilistic Event Attribution uses large ensembles of climate model simulations to compare observed events to model-generated counterfactuals
- Storyline Approach examines the physical processes driving an event and evaluates how anthropogenic forcings might have modified those processes

8.3 Attribution of global warming

Attribution statements for global warming in the three most recent IPCC Assessment reports are:

AR4: **Most of the observed increase** in global average temperatures since the mid-20th century is *very likely* due to the observed increase in anthropogenic greenhouse gas concentrations. (IPCC 2007)

AR5: It is *extremely likely* that **more than half of the observed increase** in global average surface temperature from 1951 to 2010 was caused by the anthropogenic increase in greenhouse gas concentrations and other anthropogenic forcings together. The best estimate of the human-induced contribution to warming is similar to the observed warming over this period. (IPCC 2013)

AR6: The *likely* range of total human-caused global surface temperature increase from 1850–1900 to 2010–2019 is 0.8°C to 1.3°C, with a best estimate of 1.07°C. It is *likely* that well-mixed GHGs contributed a warming of 1.0°C to 2.0°C, other human drivers (principally aerosols) contributed a cooling of 0.0°C to 0.8°C, natural drivers changed global surface temperature by –0.1°C to +0.1°C, and internal variability changed it by –0.2°C to +0.2°C. It is *very likely* that **well-mixed GHGs were the main driver** of tropospheric warming since 1979. (IPCC 2021)

The AR4 and AR5 attribution statements reference the warming since the mid-20th century, the period when greenhouse gas emissions began increasing rapidly. The words “most” and “more than half” are deliberately imprecise, potentially ranging from 51 to 99 percent of the warming – presumably this imprecision is to account for structural uncertainties such as natural internal variability. The confidence level increases from *very likely* to *extremely likely* from AR4 to AR5. The structure of the attribution statement in the AR6 is fundamentally different, referencing the warming to the period 1850-1900. The AR6 attribution statement is more precise numerically, but with a lower level of confidence at “likely” – AR6 attributes essentially all the warming to increases in greenhouse gases. The most confident statement from AR6 relates to the tropospheric warming since 1979, using the words “main driver” and “*very likely*.”

There are three areas of substantive criticism of the IPCC’s assessment of the causes of the recent warming: inadequate assessment of natural climate variability, inappropriate statistical methods, and substantial discrepancies between models and observations. The last is discussed in Chapter 5, while this chapter discusses the first two factors. All of these criticisms are relevant to the IPCC’s attribution of the recent warming, which also underpins extreme event attribution

8.3.1 Natural climate variability

AR6 states that natural external drivers since 1850-1900 have changed global surface temperature by –0.1°C to +0.1°C, and internal variability has changed it by –0.2°C to +0.2°C – on average having essentially no net impact on the warming since 1850-1900. As discussed below, this minimal contribution of natural variability has been disputed by several publications that question the magnitudes of solar variability and internal variability from large-scale ocean circulations.

Solar variability

AR5 concluded that the best estimate of radiative forcing due to Total Solar Irradiance (TSI) changes over the period 1750–2011 was very small (0.05 W/m², Myrhe *et al.*, 2014). AR6 acknowledges substantially higher values and a much larger range of estimates of changes in TSI over the last several centuries, stating that the TSI between the Maunder Minimum (1645–1715) and the second half of the twentieth century increased by 0.7–2.7 W/m², a range that includes both low and high variability TSI data

sets (Gulev, 2021). However, the recommended forcing dataset for the CMIP6 climate model simulations used in AR6 for attribution studies averages two data sets with low solar variability (Matthes, 2017).

While AR6 shows a substantially greater solar impact than does AR5, the overall impact of solar forcing on the climate was still assessed to be small compared to anthropogenic forcing. However, the impact of solar variations on the climate is uncertain and subject to substantial debate (Lockwood, 2012; Connolly *et al.*, 2021) - something that is not evident in the IPCC assessment reports.

The variations of TSI over time remains a challenging problem. Since 1978, there have been direct measurements of TSI from satellites. However, the data exhibits non-negligible inconsistencies, and interpreting any multi-decadal trends in TSI requires comparisons of observations from overlapping satellites. There are several rival composite TSI datasets that disagree as to whether TSI increased or decreased during the period 1986–96 (the ACRIM gap; see Chapter 4). Further, the satellite record of TSI is used to calibrate proxy models that infer past solar variations from sunspots and cosmogenic isotope measurements (Velasco Herrera *et al.*, 2015).

There is substantial evidence for high solar activity in the second half of the 20th century (starting in 1959) and extending into the 1990's, before a decline in the early 21st century; this period is often termed the “Modern Maximum.” (Chatzistergos *et al.*, 2023; Solanki *et al.*, 2004; Usoskin *et al.*, 2007). However, some scientists have concluded that it is not possible to be confident of any multi-decadal trend in TSI (Schmutz, 2021).

This uncertainty causes some reconstructions of TSI from 1750 to have low variability (implying a very low impact of solar variations on global mean surface temperature) whereas datasets with high TSI variability can explain more than 70 percent of the temperature variability since preindustrial times (Scafetta, 2013; Stefani, 2021). The choice of TSI satellite record used in an analysis can therefore substantially influence how much climate change is attributed to human versus natural forcings.

There is growing evidence that other aspects of solar variability, which are referred to as solar indirect effects, either amplify TSI forcing or are independent of TSI forcing. Scafetta *et al.* (2023) suggests that ~80 percent of solar influence on climate might stem from non-TSI mechanisms. There are numerous candidate processes, including solar ultraviolet changes; energetic particle precipitation; atmospheric-electric-field effect on cloud cover; cloud changes produced by solar-modulated galactic cosmic rays; large relative changes in the magnetic field; and the strength of the solar wind. Such solar indirect effects are not included in climate models, although indirect methods of estimating their impacts suggest they are significant. However, claims of non-TSI mechanisms influencing climate are uncertain and debated.

Natural variability of large-scale ocean circulations

Variations in global mean surface temperature are linked to recurrent large-scale variations in ocean circulation patterns, including the Atlantic Multidecadal Oscillation (AMO), the Pacific Decadal Oscillation (PDO) and the El Niño-Southern Oscillation (ENSO). These circulations influence ocean heat uptake and heat distribution and also influence atmospheric circulation patterns and cloud distributions. There is some debate as to whether these variations are strictly internal to the climate system, or whether this variability can have a solar/astronomical origin or can be influenced by large volcanic eruptions.

While climate models simulate the large-scale ocean circulations and internal climate variability, most models have too little amplitude compared to observations at multi-decadal frequencies and phasing out-of-sync with the observed climate (Kravtsov *et al.* 2024). Averaging multiple simulations effectively averages out the internal variations, leaving only the forced climate variability (*e.g.* CO₂ forcing). With most of the modern warming beginning in the late 1970's, the recent 50-year warming is on the same time scale as the multi-decadal oscillations of the AMO and PDO.

Here are summary statements from the AR5 and AR6 reports:

AR5: Decadal variability in the Pacific, associated with the PDO or IPO [Interdecadal Pacific Oscillation], contributes significantly to regional and global temperature trends, but the relative contributions of internal variability and external forcing are difficult to disentangle in CMIP5 simulations.

AR6: Since AR5, there has been increased understanding of the role of internal variability, such as ENSO, PDO, and AMO, in modulating regional climate trends. However, limitations in simulating the exact timing and amplitude of these modes in CMIP6 models contribute to uncertainties in attributing observed changes to anthropogenic forcing (*high confidence*).

The amplitude of the peak-to-trough impact of the multi-decadal oscillations on global temperatures has been assessed by the IPCC AR6: “The likely range of change due to internal variability is -0.2°C to $+0.2^{\circ}\text{C}$ (IPCC, 2021).” This implies a trough-to-peak change of 0.4°C . Over many centuries, any global temperature changes from the troughs and peaks will cancel out, with little to no net impact.

However, with a nominal timescale of 60-80 years for oscillations such as the AMO and PDO, the timing of the peaks and troughs can be confused with the secular trend. This becomes very relevant for attributing the warming for the past 50 years, when most of the recent warming has occurred. Even if climate models have the correct amplitude of the multidecadal oscillations, the timing of the peaks and troughs is not adequately simulated, with each model and individual ensemble simulating a different phasing. Once the simulations of the ensemble members and different models are averaged, the natural internal variability contribution is averaged out owing to differences in phasing, effectively cancelling the role of natural internal variability in the attribution process.

The time series of global surface temperature anomalies since 1850 (Figure 8.1) shows irregular variations of significant amplitude against the background of an overall warming trend and, after 1976, a pronounced difference in trends between the Northern Hemisphere and the Southern Hemisphere. The period between 1905-1945 shows a strong trend of warming. The following 30 years from 1945-1976 showed slightly declining temperatures. The most recent warming period started in 1977.

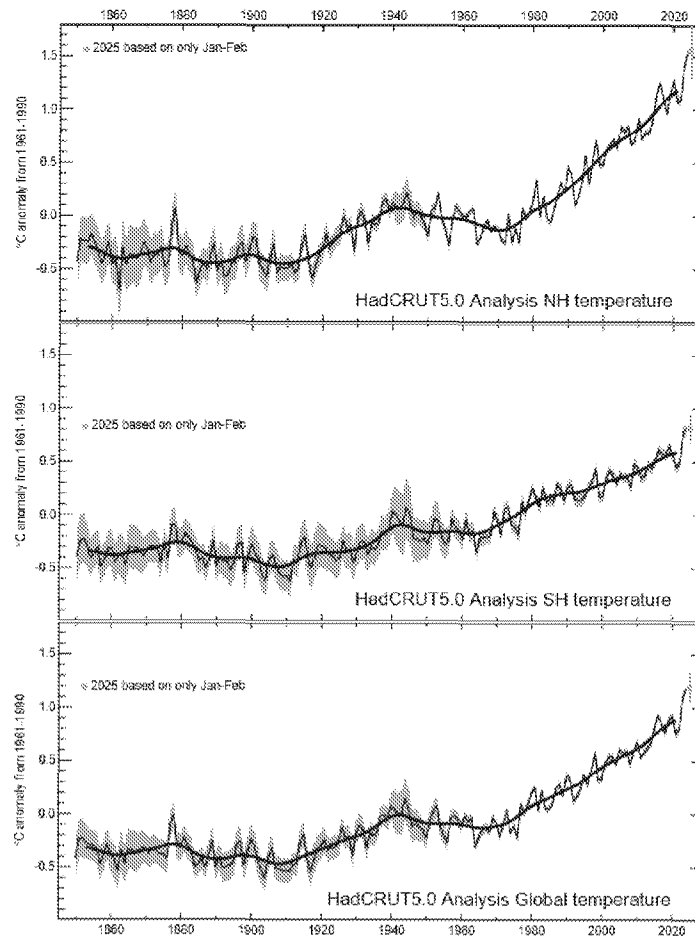


Figure 8.1 Global average surface temperature anomalies 1850—2025. Top: Northern hemisphere. Middle: Southern hemisphere. Bottom: Global average. Source: UK Hadley Centre [HYPERLINK "<https://crudata.uea.ac.uk/cru/data/temperature/HadCRUT5.0Analysis.pdf>"]

The causes of the early 20th century warming are discussed by Hegerl *et al.* (2017; 2019). The atmospheric CO₂ concentration increased from 298 ppm in 1905 to 310 ppm in 1941, implying that CO₂ had little impact. Volcanic activity during this period was very low, and solar forcing is uncertain. Yet Hegerl *et al.* (2017) somehow inferred that 40-54 percent of this warming could be attributed to external forcing, with the rest associated with internal variability. Bronniman *et al.* (2024) focused on the causes of the cooling in the first decade of the 20th century in the Southern Hemisphere. They found that the cooling was related to a La-Niña-like pattern in the Pacific, a cold tropical and subtropical South Atlantic, a cold extratropical South Pacific, and cool southern midlatitude land areas. The Southern Annular Mode was positive, with a strengthened Amundsen–Bellingshausen seas low, although the spread of the data products is considerable.

The warming in the 1930's and subsequent cooling during mid-century was particularly pronounced in the Arctic. Bokuchava and Semenov (2021) find that these variations were most likely caused by a combined effect of long-term natural climate variations in the North Atlantic and North Pacific with a contribution of the natural radiative forcing related to the reduced volcanic activity and variations of solar activity, as well as growing greenhouse gases concentrations. Tokinaga *et al.* (2017) showed that the combined effect of internally generated Pacific and Atlantic interdecadal variability intensified Arctic

warming in the early 20th century. The synchronized Pacific-Atlantic warming drastically alters planetary-scale circulations over the Northern Hemisphere; these same circulation patterns have a global influence.

The cooling period between 1945 and 1976 has been referred to as the “grand hiatus.” Numerous causes have been hypothesized: natural internal variability from fluctuations associated with the Pacific Decadal Oscillation and Atlantic Multidecadal Oscillation; cooling from increased emissions of aerosols from industrial activities; increased heat uptake in the Atlantic, Southern and Equatorial Pacific Oceans. Thompson *et al.* (2010) find that the hemispheric differences in temperature trends in the middle of the twentieth century stem largely from a rapid drop in Northern Hemisphere sea surface temperatures of about 0.3 °C between about 1968 and 1972.

The Great Pacific Climate Shift of 1976-1977 was a notable climatic event characterized by an abrupt change in the North Pacific Ocean's atmosphere-ocean system that interacted with global climate patterns. This shift is closely associated with the Pacific Decadal Oscillation (PDO) that oscillates between warm and cool phases over decades. The 1976 shift marked a transition from a predominantly negative (cool) PDO phase (1947–1976) to a positive (warm) phase (1977 through 2000), with significant implications for global and regional climate patterns, including the frequencies of El Niño and La Niña events. The Great Pacific Climate Shift coincided with the beginning of a period of accelerated global warming. When the Great Pacific Climate Shift is accounted for in climate attribution analyses since 1950, 40 percent or more of the warming in the second half of the 20th century is attributed to natural internal variability (McLean *et al.*, 2009; Tung and Zhou, 2013; Chylek *et al.*, 2016; Scafetta, 2021).

8.3.2 Optimal fingerprinting

Optimal fingerprinting is a statistical technique introduced by Allen and Tett (1999) that compares observed climate data to climate model simulations to identify patterns (or “fingerprints”) associated with human or natural forcing. It involves taking a vector of observed climate changes, such as warming rates in locations around the world, and decomposing them into a weighted sum of “signals”, which are analogues to the observations generated by climate models with different types of forcings. The weights are chosen using a regression method called Total Least Squares (TLS). The analysis commonly uses just two signals, one generated by models using only anthropogenic forcings and one using only natural forcings. If the estimated weighting coefficient attached to a signal is significantly different from zero, then that signal is said to be “detected”. If it is close to 1.0, then the model generating the signal is said to be consistent with observations. If it is less than 1.0, then the model signal for that forcing is too strong and needs to be scaled down, and vice-versa. Both the observed data and the signals are weighted using a model-generated estimate of patterns of randomness in the climate system so as, in principle, to put maximum weight on regions where natural variability is minimized.

Optimal fingerprinting is the primary tool for attribution in the research literature. While it has been widely used and prominently featured by the IPCC since 2001, there is very little literature examining the statistical properties of the results it generates. One of its inherent weaknesses is that results depend on assumptions about the accuracy of climate models, especially regarding their representation of natural variability (IPCC AR5 Ch 10.2.3).

A series of papers by McKittrick (McKittrick 2021, 2022, 2023, 2025) has challenged the optimal fingerprinting method, arguing that it is inherently unreliable and that practices adopted from econometrics can provide more valid results. Statistical theory requires that for regression models to yield unbiased coefficients, a set of assumptions called the Gauss-Markov conditions must hold. McKittrick (2021) argued that the Allen and Tett (1999) methodology violates the Gauss-Markov conditions, leading to potentially biased and inconsistent fingerprinting coefficients, undermining the reliability of the method. Chen *et al.* (2023) confirmed this analysis, although they argued that the method could yield valid results under very stringent assumptions. McKittrick (2022) and (2023) raised a further concern that climate scientists—virtually alone among

scientific disciplines—have used TLS to estimate anthropogenic greenhouse gas signal coefficients, despite its tendency to be unstable unless some strong assumptions hold that in practice are unlikely to be true. Under conditions that easily arise in optimal fingerprinting, TLS estimates have a large positive bias. Thus, any study that used TLS for optimal fingerprinting without verifying its applicability in the specific data context has likely overstated the attribution.

McKittrick (2025) presented an empirical example comparing the results of conventional optimal fingerprinting against methods drawn from mainstream econometrics that are known to be valid for the specific application of signal detection. While the IPCC optimal fingerprinting method yields an anthropogenic signal coefficient close to 1.0 on a global temperature data set spanning 1900 to 2010, the consistent method yields a coefficient around 0.4, which rises to about 0.65 on data spanning 1980 to 2010, implying the model response to greenhouse gases needs to be scaled down by about half to optimally match observations. The natural forcing signal coefficient, by contrast, is between 2.0 and 4.0, implying the climate model signals of natural forcing need to be scaled up two-to four-fold to match observed climate change. The fingerprinting coefficients estimated in McKittrick (2025), when used to scale the average sensitivity of the climate models used to generate the forcing signals in his data set, imply a Transient Climate Sensitivity of 1.4°C, which is consistent with the estimate by Lewis (2023) using a different estimation method and multiple independent data sets.

These findings indicate that the basis on which the optimal fingerprinting method has long been viewed as reliable is not valid. Re-examining previous results individually would be required to determine which findings are statistically robust.

8.3.3 Time series methods

The IPCC (AR5 WGI 10.2.2) drew attention to alternative approaches of assessing causality that emerged from the time series econometrics literature. These have the advantage of not depending on assumptions about the accuracy of climate models, but the disadvantage that they depend on difficult-to-test assumptions about the data generating process underlying climate and forcing data. The methods, now referred to as climate econometrics, use the tools of unit root testing, Granger causality and cointegration analysis, all of which are familiar in economics and finance and which are slowly being adopted in climate science. Time series analysis methods hold out the possibility of determining whether anthropogenic or natural forcings are the primary drivers of climate change without requiring the use of climate models, although key questions remain unsettled (e.g. Dergiades *et al.*, 2016, Balcombe *et al.*, 2019, Dagsvik *et al.*, 2020, Razzak, 2022, Dagsvik and Moen, 2023).

“Granger causality” modeling is a tool for determining directions of influence in data series that move together. It is a statistical, not a physical, concept. If knowing the current value of one variable significantly improves the forecast accuracy of another variable, we can infer a causal connection exists; this is called Granger causality. The modeling tools, called vector autoregression, measure direct impulse-response patterns and feedbacks. An application to the well-known Vostok ice core data revealed an error in Al Gore’s documentary *An Inconvenient Truth*. Gore showed the data and drew attention to the coherence of temperature and CO₂ changes over a 440,000 year span, which he asserted was due to CO₂ driving temperature changes. But temperature changes can also affect atmospheric CO₂ levels. Davidson *et al.* (2015) examined the series and found that temperature Granger causes CO₂ but not the reverse. In other words on the time scales represented in the Vostok data, the coherence in the series is primarily due to the influence of temperature on CO₂ levels, not the feedback of CO₂ levels on temperature.

In summary, the primary statistical methods for attributing causation in climate data are optimal fingerprinting and time series analysis. Applications of the Allen and Tett (1999) optimal fingerprinting method dominate the attribution literature and have underpinned past IPCC conclusions, but results depend

on the accuracy of climate models and the method has recently been criticized as inherently biased. Time series methods do not depend on climate models but require assumptions of their own and have generated results that have thus far not converged on a consensus.

8.4 Declining planetary albedo and recent record warmth

A sharp recent increase in global average temperatures has raised the question of short-term drivers of climate. One such candidate is the fraction of absorbed solar radiation which has also increased abruptly in recent years. The question is whether the change is an internal feedback to warming caused by greenhouse gases, or whether something else increased the fraction of absorbed radiation which then caused the recent warming.

The planetary albedo is the fraction of incoming solar radiation that is reflected back into space rather than being absorbed by the planet. Highly reflective surfaces like cloud tops and snow and ice are most important in this regard. The Earth's albedo is approximately 30 percent, meaning almost a third of the sunlight that reaches Earth is directly reflected back to space. A lower albedo implies more solar energy is absorbed by the planet to be then re-radiated as heat. Hence, other things being equal, a decline in planetary albedo is associated with a warming of the Earth.

Arguably the most striking change in the Earth's climate system during the 21st century is a significant reduction in planetary albedo since 2015, which has coincided with at least two years of record global warmth. Figure 8.2 shows the planetary albedo variations since 2000, when there are good satellite observations. The 0.5 percent reduction in planetary albedo since 2015 corresponds to an increase of 1.7 W/m² in absorbed solar radiation averaged over the planet (Hansen and Karecha, 2025). For comparison, Forster et al. (2024) estimate the current forcing from the increase in atmospheric CO₂ compared to preindustrial times to be 2.33 W/m².

Looking back prior to 2000 with less adequate data, Goessling *et al.* (2025) assessed that the planetary albedo was relatively low around the 1940's and 50's before rising industrial aerosol precursor emissions increased the albedo until the 1980's. The strongest planetary albedo excursions were high-albedo episodes caused by volcanic eruptions, such as after the Mount Pinatubo eruption in 1991. Although uncertain, the 2023 planetary albedo minimum might have been the lowest since at least 1940.

Changes in surface characteristics cannot explain this decrease in planetary albedo since 2015:

- Arctic sea ice extent has declined by about 5 percent since 1980 (https://nsidc.org/data/seaice_index/images/s_plot_hires.png), although following 2007 there has been a pause in the Arctic sea ice decline (England *et al.*, 2025)
- Regarding Antarctic sea ice, the IPCC AR6 concludes that "There has been no significant trend in Antarctic sea ice area from 1979 to 2020 due to regionally opposing trends and large internal variability." (Summary for Policymakers, A.1.5)
- Northern hemispheric annual snow cover has been slowly declining since 1967, with barely significant trends. The data show the Northern Hemisphere has snowier winters, accompanied by more rapid melt in spring and summer, see [[HYPERLINK "http://climate.rutgers.edu/snowcover/"](http://climate.rutgers.edu/snowcover/)] and Section 5.6
- Global greening (Chapter 2) is contributing to the decrease in planetary albedo, as forests have a lower albedo than open lands or snow. However, there is some evidence that forests increase cloud cover (high reflectivity), which counteracts the direct albedo decrease associated with increasing forested area.

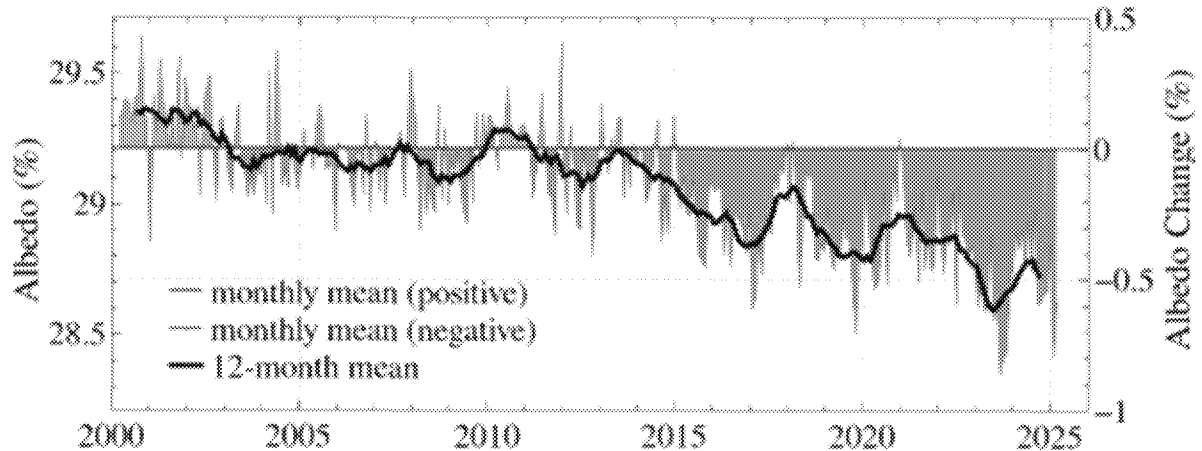


Figure 8.2. Earth's albedo (reflectivity, in percent), with seasonality removed. From Hansen and Karecha (2025)

Changes in surface reflectivity are contributing to a small, slow decline in planetary albedo. However, these changes cannot explain the sharp decline in planetary albedo beginning in 2015. Further, any changes in surface albedo are effectively damped by about a factor of three on average, primarily due to cloud masking (Loeb *et al.* 2019). Surface albedo changes have thus contributed only weakly to the recent planetary albedo decline, particularly when averaged annually and globally. This leaves the most likely explanation for the sharp albedo decline as changes in atmospheric aerosols and/or clouds.

Aerosols are small particles in the atmosphere that reflect sunlight and can interact with cloud processes to make the clouds more reflective. Aerosols have both natural origins (*e.g.* dust from soils, wildfires) as well as human origins from combustion (including fossil fuels). Regions that are greening and/or have an increase in rainfall might produce less dust. For anthropogenic aerosols, new ship fuel regulations aimed at reducing sulfur emissions were implemented in three phases, in 2010, 2015 and 2020. Sulfate particles reflect solar radiation, so cleaner air means less solar radiation is reflected, which contributes to surface warming. Indirect effects of sulfate aerosols include increasing the reflectivity of low-level clouds in the subtropics. There is controversy surrounding the importance of this change in sulfate emissions to the Earth's radiation balance (Hodnebrog *et al.*, 2024; Yuan *et al.*, 2024), but the restriction of this effect to major shipping routes suggests a small global impact (Schmidt, 2024).

Changes in clouds are the primary candidate to explain the decline in global albedo since 2015. Two recent papers (Loeb *et al.*, 2024; Goessling *et al.* 2024) have addressed recent variations in cloud properties. By considering satellite and reanalysis data, Loeb *et al.* found that decreases in low- and mid-level clouds since 2015 are the primary reason for decreasing planetary albedo in the Northern hemisphere, whereas in the Southern hemisphere the decrease in planetary albedo is primarily due to decreases in mid-level clouds across all latitude zones. Goessling *et al.* (2024) found that cloud anomalies were mainly due to reduced low-level clouds. Regions with coherent low-level cloud reductions over the past decade include the warm pool region around the Maritime Continent and the northern extratropical western Pacific, as well as large parts of the Atlantic and adjacent land regions. The reduction of global cloud cover identified in these analyses since 2015 is 1-2 percent.

The issue then becomes the cause of the change in cloud cover. Two explanations have been posited for the declining cloud cover over the past decade:

- Natural climate variability

- Changes in low cloud cover associated with warming sea surface temperatures, implying an emerging positive feedback to climate change (Hansen and Karecha, 2025)

It is not easy to justify a new positive low cloud feedback that began emerging in 2015 since there is no obvious feedback trigger starting at that time. However, there are numerous natural climate signals during this period that are associated with atmospheric circulation changes that can influence the distribution of clouds:

- The 2014-2016 was one of the strongest El Niño events on record
- A cold anomaly beginning in 2015 in the subpolar gyre of the North Atlantic reflects a shift in the ocean circulation pattern associated with decadal variability in the Atlantic (Frajka-Williams *et al.*, 2017; Arthun *et al.* 2021).
- The Pacific Decadal Oscillation positive index peaked in 2016, then declined and has been in negative territory since late 2019
- Eruption of the submarine Hunga-Tonga volcano in 2022

Interannual cloud anomalies associated with the El Niño Southern Oscillation (ENSO) have a significant global signal and strong regional signals, especially over the tropical Indian and Pacific Oceans.

The Hunga Tonga eruption is remarkable in its coincidence with the exceptionally low value of planetary albedo in 2023. Goessling *et al.* (2024) found that the cloud perturbations in 2023 showed a different pattern from the overall pattern since 2015, with reduced cloud cover being most pronounced in the northern hemisphere and tropics. Regional reductions in cloud cover were most pronounced over the eastern Indian Ocean, over South America and extending over the eastern Pacific, as well as over northern North America, in the Southern Ocean around 60S, in the subtropical and eastern North Atlantic, and in parts of the North Pacific.

The Hunga Tonga eruption was unusual in that it injected large amounts of water vapor into the stratosphere in addition to sulfate particles. Early publications focused on the impacts to stratospheric circulations and the direct radiative impact. In 2025, papers are emerging that examine the indirect effects of Hunga Tonga on surface climate using ensemble simulations from earth system models (Bednarz *et al.*, 2025; Kuchar *et al.*, 2025). These papers found statistically significant impacts on regional climates that were driven by circulation changes from couplings between the stratosphere and troposphere. These papers indicate the complex interactions between volcanic activity and climate dynamics. More research is needed to untangle any impacts of the Hunga Tonga eruption on the planetary albedo.

In summary, the decline in planetary albedo and the concurrent decline in cloudiness have emphasized the importance of clouds and their variations to global climate variability and change. A change of 1- 2 percent in global cloud cover has a greater radiative impact on the climate than the direct radiative effect of doubling CO₂. While it is difficult to untangle causes of the recent trend, the competing explanations for the cause of the declining cloud cover have substantial implications for assessing the Equilibrium Climate Sensitivity and for the attribution of the recent warming. An additional 10 years of data should help clarify whether this is a strong positive cloud feedback associated with warming or a temporary fluctuation driven by natural variability

8.5 Attribution of climate impact drivers

The IPCC (Ranasinghe *et al.* 2021) defines “climate impact drivers” or CIDs as “physical climate system conditions (e.g., means, events, extremes) that affect an element of society or ecosystems.” Hence CIDs are those features of the weather and climate system of primary interest in assessing the impacts of climate change since they potentially affect humans and the natural world. For instance, under the heading “Heat and Cold,” CIDs are identified as mean air temperature, extreme heat, cold spells and frost. The IPCC

also points out that CIDs are not necessarily harm-related: depending on the system in question they can be detrimental, neutral, beneficial or a combination.

The first columns Table 12.12 in Ranasinghe *et al.* (2021, p. 1856), reproduced here as Table 8.1, summarize the AR6 assessment of whether a signal of attributable anthropogenic influence has emerged across all major CIDs. One of the themes of this chapter is that attribution methods used by the IPCC tend to overstate the anthropogenic influence and understate the role of natural variability. Nonetheless a striking feature of that summary table is how few CIDs exhibit an anthropogenic signal sufficient to distinguish them from natural variability.

Out of the 33 weather impact categories listed, an anthropogenic signal is asserted with *high confidence* in only five, and with *medium confidence* in a further four. (Note that one of the CIDs is an increase in CO₂ levels, and since it is a tautology to attribute this to increased CO₂ levels this CID can be ignored.) For the rest the IPCC does not claim to have detected anthropogenic drivers. Of the five *high confidence* assertions, two are for changes in average temperatures (air and ocean) hence are not measures of extreme weather. Further, two of the four *medium confidence* assertions are related to ocean chemistry and thus are likewise not related to extreme weather. The IPCC does not assert a human influence on many non-temperature weather features such as wind, precipitation, flooding, or drought.

Other columns of IPCC Table 12.12 report on whether anthropogenic signals are expected to emerge this century under RCP8.5, the most extreme forcing scenario. We have omitted these columns for several reasons. First, because they refer to whether climate models project detectable signals in the observations, which is a very different question than our concern here: whether a signal has been detected in historical data. Second, as we discuss in Chapter 4, the RCP8.5 scenario is a misleading and implausible high-end storyline, it is not a “base case” or business-as-usual projection. Third, the ensemble of detailed models is acknowledged (Palmer and Stevens 2019) to be “not fit for purpose” in describing regional changes and “unable to represent future conditions at the degree of spatial, temporal, and probabilistic precision with which projections are often provided.” (Nissan et al, 2019). Even with these caveats, we note that the omitted columns in AR6 Table 12.12 show most weather impacts are not expected to exhibit an anthropogenic signal through the end of this century.

As discussed in Chapter 6 natural variability dominates patterns of extreme weather systems and simplistic assertions of trend detection are frequently undermined by regional heterogeneity and trend reversals over time. Table 8.1 makes the related point that it is not currently possible to attribute changes in most extreme weather types to human influences. Taking wind as an example, the IPCC claims that an anthropogenic signal has not emerged in average wind speeds, severe windstorms, tropical cyclones or sand and dust storms, nor is one expected to emerge this century even under an extreme emissions scenario. The same applies to drought and fire weather.

Climatic Impact-driver Type	Climatic Impact-driver Category	Already Emerged in Historical Period
Heat and Cold	Mean air temperature	1
	Extreme heat	2
	Cold spell	4
	Frost	
Wet and Dry	Mean precipitation	
	River flood	
	Heavy precipitation and pluvial flood	
	Landslide	
	Aridity	
	Hydrological drought	
	Agricultural and ecological drought	
	Fire weather	
Wind	Mean wind speed	
	Severe wind storm	
	Tropical cyclone	
	Sand and dust storm	
Snow and Ice	Snow, glacier and ice sheet	
	Permafrost	
	Lake, river and sea ice	11
	Heavy snowfall and ice storm	
	Hail	
	Snow avalanche	
Coastal	Relative sea level	
	Coastal flood	
	Coastal erosion	
Open Ocean	Mean ocean temperature	
	Marine heatwave	
	Ocean acidity	
	Ocean salinity	13
	Dissolved oxygen	14
Other	Air pollution weather	
	Atmospheric CO ₂ at surface	
	Radiation at surface	

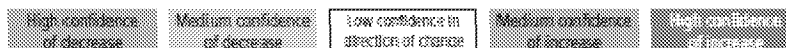


Table 8.1: Reproduction of column 1 of Table 12.12, IPCC AR6 Working Group I report. Emergence of anthropogenic signal in historical period for CIDs shown. White: no signal detected. Blue and orange: change detected (decrease or increase) and confidence level as indicated in color legend. Numbers refer to specific regions and confidence levels: see original Table for notes.

8.6 Extreme event attribution (EEA)

Examining the IPCC's treatment of extreme weather and climate event attribution more broadly, AR6 provides an ambiguous assessment of the role of anthropogenic warming that differs between its chapters. Chapter 11 of WG1 states (Seneviratne *et al.*, 2021):

Evidence of observed changes in extremes and their attribution to human influence (including greenhouse gas and aerosol emissions and land-use changes) has strengthened since AR5, in particular for extreme precipitation, droughts, tropical cyclones and compound extremes (including dry/hot events and fire weather). Some recent hot extreme events would have been extremely unlikely to occur without human influence on the climate system.

By contrast, as noted in Section 8.5, Chapter 12 of WG1 (Table 12.12) paints a different picture – presumably, the expert judgment of different groups of authors for the two chapters came to different conclusions (Ranasinghe, 2021):

- High confidence in an increase in extreme heat events in tropical regions where observations allow trend estimation and in most regions in the mid-latitudes, medium confidence elsewhere
- Medium confidence in a decrease in extreme cold events in Australia, Africa and most of northern South America where observations allow trend estimation
- No evidence of emergence in the historical period of a change in river floods, heavy precipitation, drought, fire weather, severe windstorms, and tropical cyclones

While the overall issue of detecting changes in extreme weather events and their attribution remains ambiguous, most of the activity in this area relates to the attribution of particular extreme weather events. The most prominent effort is World Weather Attribution (WWA; worldweatherattribution.org), an international research initiative for extreme event attribution that purports to analyze how climate change influences the likelihood and intensity of extreme weather events. Their approach is to use large ensembles of regional climate models to compare an event in today's climate with one in a counterfactual pre-industrial climate without human influences.

WWA has a prominent public presence in linking extreme weather to climate change, with its press releases attracting considerable attention in public and policy discussions. However, WWA's extensive promotion of non-peer-reviewed findings, its open admission to shaping analyses to serve litigation, and its methodological challenges have sparked controversies, with critics questioning the robustness and impartiality of their conclusions (Pielke Jr. 2024). Despite these issues, WWA's work continues to influence climate science and media narratives. Technical criticisms of the approach include a lack of a formal detection process; an implicit assumption that 100 percent of the post-industrial warming is caused by greenhouse gases; and a failure to adequately account for internal climate variability.

Because EEA is relatively new, many basic methodological issues have yet to be settled in the expert literature. An important challenge is the lack of data. Extreme events are by definition rare. Many analyses of extreme event types (including the U.S. National Assessment Reports) only evaluate data since 1950 or 1970. However, as emphasized in Chapter 6, many of the worst extreme weather and climate events in U.S. history occurred in or before the first half of the 20th century, including in the early 19th century. And if paleoclimate reconstructions are considered, it becomes very difficult for an event to pass thresholds of what is expected from natural variability, particularly if a reasonably sized geographic region is considered.

Another challenge is defining the event under study. There is a longstanding literature in statistics and econometrics on the challenge of analyzing data with outliers. The issue arises because a data series establishes a probability distribution defining the expected range of observations. If an outlier is observed it might indicate that the underlying process giving rise to the data distribution has changed (which in the weather context would mean that a climate change has been detected) or that the underlying process has multiple regimes each with a different probability distribution, in which case observing an outlier simply means we were temporarily in a different regime, but the system itself was unchanged. If a time series contains only a single outlier event at the end of the series, it is not possible to determine which model is the correct one (Chen and Liu 1993). For instance, there might be an “ordinary” weather regime that yields a distribution of summer daytime highs in a particular coastal region, and a second “heatwave” regime that kicks in when an inland blocking event occurs, which yields a temperature distribution centered 15°C higher than the first one. A day with temperatures 13°C above normal would either be an extreme heat anomaly under the first regime or a somewhat cool event under the second, and we have no way in this case of knowing on statistical grounds which view is correct.

Visser and Petersen (2012) and Sardeshmukh *et al.* (2015) both point out that different distributions might fit observed data equally well but yield very different implications about the likelihood of a specific weather event. Visser and Petersen argue that, in view of the deep uncertainties of extreme weather analysis, drawing a connection between individual events and global climate change should be avoided. Furthermore, the existence of an outlier at the end of a data series poses the problem that estimates of the event probabilities will be biased whether the outlier is included or excluded (Barlow *et al.*, 2020). Methods to eliminate the bias have not yet been established, leading some experts (*e.g.* Miralles and Davison 2023) to argue that in settings in which a data series contains a single extreme event at the end, estimation of a return period for the extreme event will be so biased and uncertain that it should be avoided altogether.

We provide a case study of a recent high impact extreme event in the U.S. to illustrate the challenges and ambiguities in attributing the frequency and intensity of extreme weather events to human-caused warming.

8.6.1 Case study – 2021 Western North America heat wave

The 2021 Western North America heat wave was an extreme event that affected much of Western North America in late June 2021. Surface temperature records were set in Portland, OR (116°F; previous record 107°F) and Seattle, WA (108 °F; previous record 103°F) (Mass *et al.*, 2024).

The WWA team generated international headlines with their analysis, which provided the following attribution statements (WWA, 2021; Philip *et al.*, 2022):

- Based on observations and modeling, the occurrence of this heatwave was virtually impossible without human-caused climate change.
- The event is estimated to be about a one in 1000-year event in today’s climate.
- The event would have been at least 150 times rarer without human-induced climate change.
- This heatwave was about 2°C hotter than it would have been if it had occurred at the beginning of the industrial revolution (when global mean temperatures were 1.2°C cooler than today).

But an important counter to the first claim is that other researchers concluded from historical weather data that while a heat wave of the magnitude observed was indeed virtually impossible without anthropogenic climate change, it was also virtually impossible *with* climate change. Bercos-Hickey (2022) noted “these temperatures were virtually impossible under any previously experienced meteorological conditions, with or without global warming.” McKinnon and Simpson (2022) stated “the most likely explanation remains that the weather event itself was ‘bad luck.’” The 2023 Oregon Climate Assessment (Fleischman, 2023) concluded that the heat dome would have formed even without climate change and “There is no evidence that the highly unusual combination of weather features that drove the heat dome

were made more likely by climate change, and climate models do not project an increase in the frequency of high-pressure ridges over the Pacific Northwest” (Fleischman, 2023, p. 49).

Mass *et al.* (2024) summarizes the proximate sequence of compound events leading to the heat wave. There was a record-breaking mid-tropospheric ridge over the Pacific Northwest, forced by a tropical disturbance in the western Pacific. This produced record-breaking mid-tropospheric temperatures, strong subsidence in the lower atmosphere, low-level easterly flow that produced downslope warming on regional terrain and the removal of cooler marine air, and an approaching low-level trough that enhanced downslope flow. The event occurred at a time of maximum solar insolation, and drier-than-normal soil moisture. Using a storyline approach, Mass *et al.* assessed that there was no trend in drought and dry soils in the Pacific Northwest; there is no evidence of global warming producing stronger ridges of high pressure, and no observed trend in heat waves or record temperatures in the region. They concluded that although anthropogenic warming might have contributed as much as 2°F to the magnitude of the event, there is little evidence of further amplification of the event from increasing greenhouse gases.

Bercos-Hickey *et al.* (2022) conducted a statistical analysis of a model-based attribution study of this heat wave. Because the event is a far outlier and far above the bounds of Generalized Extreme Value distributions fitted from historical data, they concluded that estimates of return times, quantitative changes in event magnitude and frequency, and probability of the extreme temperatures such as provided by the WWA are not accurate and should be interpreted as having low confidence. They found that hindcast attribution methods using an ensemble of regional climate models, combined with Pearl (2009) causal inferences, can provide limited and conditional information about the magnitude of the human influence on this heatwave - they provided a more highly constrained estimate that human activities caused a ~1.4°F–1.8°F increase in the daily maximum temperatures.

Zeder *et al.* (2023) also concluded that the methods employed by Philip *et al.* (2022, the WWA analysis) tend to overstate the rarity of extreme heat waves, leading to a biased perception of the effect of climate change on the heatwave event: “The tendency to overestimate the return period of observed extreme heatwave events may fuel the impression that seemingly impossible heatwave extremes are currently clustering at an unprecedented rate.”

The 2021 Western North America heatwave was a rare and unprecedented compound weather event that broke century-old temperature records by as much as 9°F. While the WWA team received worldwide publicity for their rapid attribution claims blaming anthropogenic climate change, subsequent peer-reviewed analyses showed that the event was caused by rare meteorological conditions that were not made more probable by global warming.

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PART III: IMPACTS ON ECOSYSTEMS AND SOCIETY

9 CLIMATE CHANGE AND U.S. AGRICULTURE

Chapter summary

There has been abundant evidence going back decades that rising CO₂ levels benefit plants, including agricultural crops, and that CO₂-induced warming will be a net benefit to U.S. agriculture. The increase in ambient CO₂ has also boosted productivity of all major U.S. crop types. There is reason to conclude that on balance climate change has been and will continue to be neutral or beneficial for most U.S. agriculture.

9.1 Econometric analyses

Econometric analyses of the effects of climate change on agriculture have sought to integrate information about long term crop yield changes in response to temperature and precipitation changes under assumptions about adaptive behavior by farmers. One method focuses on variations in agricultural land values. The rationale is that if climate change is a long-term net benefit for agriculture it should be capitalized into higher market values for agricultural land, and vice versa. While individual crops might benefit from, or be harmed by, climate change, once the adaptive responses of farmers are considered the value of farmland represents an index of whether the changes are expected to be beneficial or not. Mendelsohn *et al.* (1994) examined the relationship between historical climatic variations on agricultural land values and concluded global warming would be slightly beneficial to American agriculture.

The Mendelsohn *et al.* method, called Ricardian analysis after David Ricardo, the 19th century British economist who pioneered the study of land values, attracted subsequent criticism from authors who argued it failed to account for differences in land values attributable to fixed locational characteristics like soil quality, and non-climatic changes including nearby urbanization. Deschênes and Greenstone (2007) looked at agricultural profits instead of land values and reached conclusions similar to Mendelsohn *et al.* (1994), namely that past climatic variations had relatively little effect on farm profitability and that warming would likely yield small overall benefits for the U.S. agricultural sector. However, a subsequent exchange with critics led them (Deschênes and Greenstone 2012) to revise their conclusions and project potentially large losses in U.S. agriculture due to climate warming.

Burke and Emerick (2016) looked at temperature variations over 1980 to 2000 and argued that farmers were not as able to adapt to temperature changes as the Ricardian method assumes and that climate change would have large negative impacts on corn and soy yields. Schlenker and Roberts (2009) similarly argued that yield gains to past warming would not carry over to the future and corn and soy yields would sharply decrease this century due to climate change.

Two recent studies have argued that pessimistic findings such as these are not robust. Ortiz-Bobea (2019) argued that land values aggregate farm and non-farm influences and the latter need to be filtered out. He developed a data set using cash rents for agricultural activity as a measure of land value specifically for farming activity. Whereas the land value model implied future losses under climate warming, the same model estimated using cash rents did not, leading the author to conclude the pessimistic results were due to using an inaccurate measure of the returns to farming activity. Bareille and Chakir (2023) assembled a large data base on farm sale prices in France for properties that sold twice between 1996 and 2019. They could replicate pessimistic results showing negative effects of warming on agricultural land values using conventional econometric modeling. But by taking advantage of the repeat sales data, which provides information on site-specific changes in land prices, they found the results reversed and implied that climate change will be very beneficial for French agriculture. The authors concluded that, taking adaptation into account, a warming climate would yield positive benefits for French agriculture that were between two and

20 times larger than had previously been estimated. On average, with full adaptation, they concluded that climate changes under the medium RCP4.5 scenario could double the value of French farmland by 2100.

A major deficiency of all these studies, however, is that they omit the role of CO₂ fertilization. Climate change as it relates to this report is caused by GHG emissions, chiefly CO₂. The econometric analyses referenced above focus only on temperature and precipitation changes and do not take account of the beneficial growth effect of the additional CO₂ that drives them. As explained in Chapter 2, CO₂ is a major driver of plant growth, so this omission biases the analysis towards underestimation of the benefits of climate change to agriculture.

9.2 Field and laboratory studies of CO₂ enrichment

One of the ways the effect of CO₂ on crop growth has been studied is through “free air enrichment experiments” or FACE plots, in which small sources of CO₂ are placed in fields surrounding plants and the growth response to elevated CO₂ under varying weather conditions are recorded. Ainsworth *et al.* (2020) summarizes results from about 250 such studies. They found that elevation of CO₂ by 200 ppm caused an average 18 percent increase in crop yield in C3 plants. C4 plants exhibited benefits mainly under drought conditions.

In addition to FACE plot experiments there have been thousands of laboratory experiments on the effects of CO₂ on all kinds of plant growth. Here we review some of the results on key U.S. agricultural crops.

Soybean

Studies on the impact of elevated CO₂ on Soybean (*Glycine max* (L.) Merr.) plants in the water-deficient region of Huang-Huai-Hai Plain, China, showed that elevated CO₂ concentrations improved photosynthesis rate, water use efficiency, and growth Li (2013) under both normal conditions and drought conditions. The CO2Science.org website reports on 108 published experiments between 1985 and 2019 exposing soybean to enriched CO₂ levels. Converted to a +300ppm common scale the average growth benefit was +50.9 percent. There were also ten studies reporting on +600 ppm CO₂ enrichment, which increased photosynthesis by an average of 90.3%.

Maize (corn)

The CO2Science.org website reports on 28 published experiments between 1983 and 2018 exposing corn (*Zea mays* L.) to enriched CO₂ levels. Converted to a +300ppm common scale the average growth benefit was +23.7 percent. Corn also benefits from increased drought tolerance under elevated CO₂. An experimental study (Allen Jr., 2011) exposed plants to water stress conditions in sunlit controlled-environment chambers at 360 ppm (ambient) and 720 ppm (elevated) CO₂. The drought stress caused a 41 percent loss of growth under ambient CO₂ but only a 13 percent loss under elevated CO₂.

Wheat

The CO2Science.org website reports on 92 published experiments between 1983 and 2020 exposing common wheat (*Triticum aestivum* L.) to enriched CO₂ levels. Converted to a +300ppm common scale the average growth benefit was +67.6 percent. Blandino (2020) measured both the yield and nutritional quality of an “improver” hybrid wheat variety and its parents under elevated CO₂ levels (+166 ppm). They reported a grain yield increase of +16 percent but a 7 percent decrease in grain protein levels. But they also found that the food quality of different wheat varieties responded differently to elevated CO₂ levels, so that with proper varietal selection, growers could select wheat types that best take advantage of the elevated CO₂.

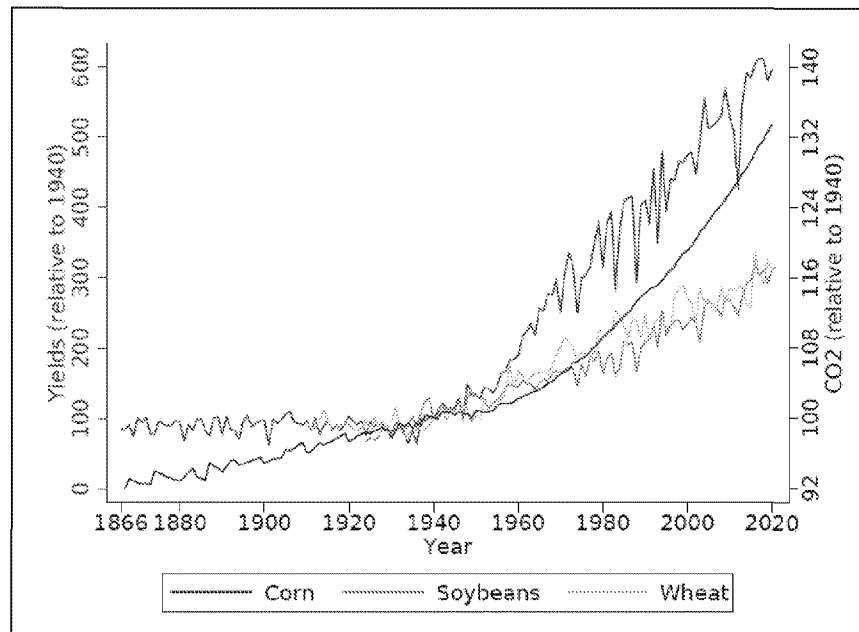


Figure 9.1: U.S. average CO₂ levels and yields of corn, soy and wheat all normalized so 1940=100. Source: Taylor and Schlenker (2021)

Further evidence

A 2021 report from the U.S. National Bureau of Economic Research (Taylor and Schlenker 2021) used satellite-measured observations of outdoor CO₂ levels across the United States, matched to county-level agricultural output data and other economic variables. After controlling for the effects of weather, pollution and technology the authors concluded that CO₂ emissions had boosted U.S. crop production since 1940 by 50 to 80 percent, attributing much larger gains than had previously been estimated using FACE experiments. They found that every ppm of increase in CO₂ concentration boosts corn yields by 0.5 percent, soybeans by 0.6 percent, and wheat by 0.8 percent.

Beyond growth benefits, extra CO₂ boosts plant resilience to dryness. See discussion in Section 2.1.3.

9.3 Crop modeling meta-analyses

Notwithstanding the abundant evidence for the direct benefits of CO₂ and of CO₂-induced warming on crop growth, in 2023 the U.S. Environmental Protection Agency (EPA 2023) boosted its estimate of the Social Cost of Carbon (SCC) about five-fold based largely on a very pessimistic 2017 estimate of global agricultural damages from climate warming (Moore *et al.*, 2017). One of the two damage models used by the EPA attributed nearly half of the 2030 SCC to projected global agricultural damages based on the Moore *et al.* (2017) analysis. This study was a meta-analysis of crop model studies simulating yield changes for agricultural crops under various climate warming scenarios. Moore *et al.* projected declining global crop yields for all crop types in all regions due to warming.

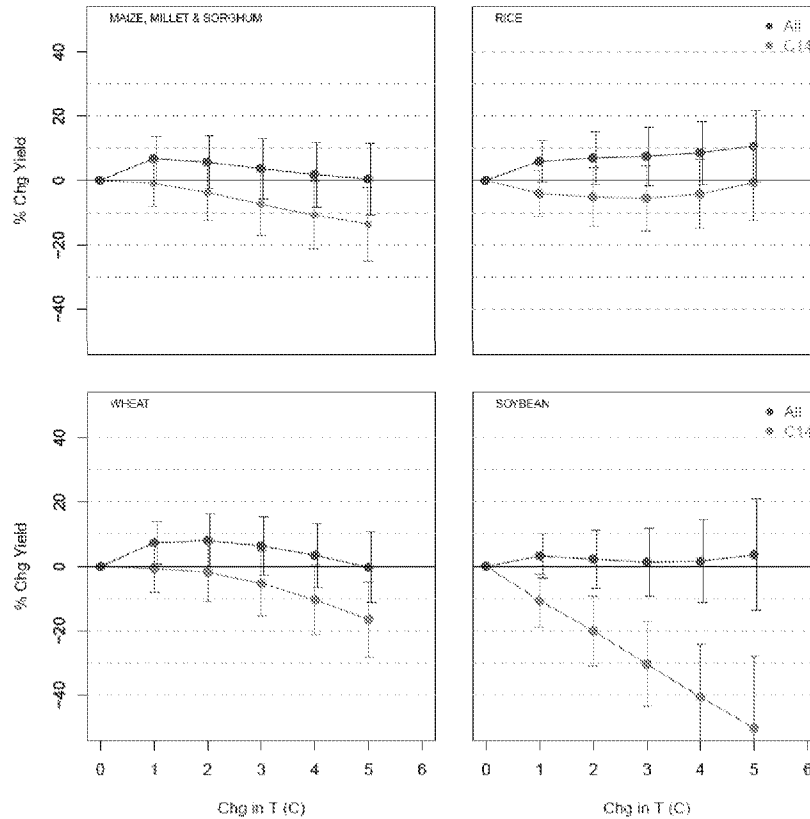


Figure 9.2 Crop yields under CO₂-induced climate warming. Blue: as published in Moore *et al.* (2017). Green: after including omitted data. Source: McKittrick (2025).

McKittrick (2025) re-examined the Moore *et al.* database and found that, while it claimed to cover 1,722 studies, only half the entries ($N=862$) had complete records, so that the sample available for regression analysis was much smaller than both studies indicated. McKittrick noted that the records most commonly missing were the changes in ambient CO₂ and found that in many cases these could be recovered from the underlying studies or the original climate scenario tables, thereby increasing the usable sample size by 40 percent. The crop yield projections incorporating the newly available data changed considerably. As shown in Figure 9.2, whereas the partial data set implied warming would decrease yield (blue lines), the complete data set implied constant or increase global yields, even out to 5°C warming (green lines).

9.4 CO₂ fertilization and nutrient loss

Evidence has shown that CO₂-induced biomass gains are sometimes accompanied by reductions in the concentrations of protein and other key nutrients such as iron and zinc (Ebi *et al.* 2021). Some experiments have shown that the rising temperatures expected to accompany higher CO₂ levels will offset this loss (Köhler *et al.* 2019) although the evidence for this is mixed, as is the evidence that nutrient dilution observed to date is entirely attributable to higher CO₂ (Ziska 2022). If nutrient dilution does occur under rising CO₂ levels, there are several adaptive strategies that could be pursued.

First, selective breeding to raise micronutrient content is already established (Saltzman *et al.* 2017) and has proven to be a cost-effective agronomic strategy (Ebi *et al.* 2021). Strategies can include both conventional breeding and genetically-modified organisms. An example of the latter is Golden Rice, which

contains elevated levels of beta-carotene to boost biosynthesis of vitamin A in the human body. Optimal strategies will be location-specific because they vary by crop, climate and soil type (Ebi *et al.* 2021). Second, fortification of food products with micronutrients is already routine. Folic acid (a B vitamin) is added to flour and many other foods; iodine is added to table salt, most commercial breakfast cereals are fortified with iron and numerous vitamins, *etc.* Third, dietary supplements in the form of multivitamin tablets are inexpensive, widely-available and routinely consumed.

One concern about reliance on adaptive strategies is whether they are feasible in low-income countries. Micronutrient deficiency is already a problem in the developing world and dietary supplements have proven to be an effective low-cost response (Ebi *et al.* 2021). It should also be noted the IPCC emission scenarios that generate high levels of warming also involve strong income growth. The SSP scenarios³ assume that, compared to 2005 levels, global per capita income will double by 2100 in the lowest growth case (SSP3), and in the highest emission case (SSP5) global per-capita income will grow nearly 16-fold. In that scenario even the poorest regions (Africa and the Middle East) end up with a per capita income of about US\$126,000, 70 percent higher than current U.S. per capita income (about US\$75,000). Consequently the same scenarios in which CO₂ levels increase the most are also those in which global poverty is largely eliminated, in which case all countries would be able to afford dietary supplements as necessary to address micronutrient deficiencies, if they arise and cannot be addressed using on-farm agricultural strategies.

In summary, there is abundant evidence going back decades that rising CO₂ levels benefit plants, including agricultural crops, and that CO₂-induced warming will be a net benefit to U.S. agriculture. To the extent nutrient dilution occurs there are mitigating strategies available that will need to be researched and adapted to local conditions.

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³ See compilation at [HYPERLINK "<https://ourworldindata.org/explorers/ipcc-scenarios?Metric=GDP&Sub-metric=Total+radiative+forcing&Region=Global&country=SSP1+-+Baseline~SSP2+-+Baseline~SSP3+-+Baseline~SSP4+-+Baseline~SSP5+-+Baseline>"].

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10 MANAGING RISKS OF EXTREME WEATHER

Chapter summary

Trends in losses from extreme weather and climate events are dominated by population increases and economic growth. Technological advances such as improved weather forecasting and early warning systems have substantially reduced losses from extreme weather events. Better building codes, flood defenses, and disaster response mechanisms have lowered economic losses relative to GDP. The U.S. economy's expansion has diluted the relative impact of disaster costs, as seen in the comparison of historical and modern GDP percentages. Heat-related mortality risk has dropped substantially due to adaptive measures including the adoption of air conditioning, which relies on the availability of affordable energy. U.S. mortality risks even under extreme warming scenarios are not projected to increase if people are able to undertake adaptive responses.

10.1 Socioeconomic context

Risks from human-caused climate change are affected by natural weather and climate variability and are dominated by the exposure of wealth in coastal and other disaster-prone regions and vulnerabilities of poorer populations. The evolution of climate risk in the U.S. has been dominated by societal factors, rather than by changes to the actual weather and climate hazards. Deaths from weather disasters have decreased substantially since 1900, even as U.S. population grew from 76 million in 1900 to over 331 million in 2020 (Goklany 2011). For example, the Galveston hurricane killed over 8,000 people in 1900 (0.01 percent of the U.S. population), whereas the worst recent disaster, Hurricane Katrina in 2005, killed 1800 people (or 0.0006 percent of the U.S. population) (NOAA National Hurricane Center 2025; U.S. Census Bureau 2025).

Technological advances have substantially reduced losses from extreme weather events. Early warning systems, satellite monitoring, and improved weather forecasting have reduced deaths, although exact numbers are hard to quantify (Deryugina and Hsiang 2023). U.S. weather forecasting has been estimated to reduce losses from weather events with an annualized benefit of \$31.5B, protecting lives, property, and supporting agriculture and transportation (NRC 2010). Improved hurricane forecasts have reduced pre- and post-landfall spending, with annual per hurricane cost reductions estimated at \$5B (Molina and Rudik 2024).

Infrastructure improvements have contributed to substantial reductions in losses from extreme weather events. Building codes, such as those implemented in Florida after Hurricane Andrew (1992), have reduced losses by ensuring structures can withstand high winds and floods. Homes built after 2002 showed minimal damage during Hurricane Michael (2018), unlike older homes (FEMA 2020). Sea walls, like the Galveston Seawall, protect against wave action and storm surge. The New Orleans Hurricane and Storm Damage Risk Reduction System successfully mitigated storm surge during Hurricane Isaac (2012) (Battelle Memorial Institute 2013). Inland dams in the U.S. help control flooding by storing excess water during heavy rains. It is estimated that the Tennessee Valley Authority (TVA) dams prevent about \$309M in annual flood damage in the TVA region and along the Ohio and Mississippi Rivers (TVA, 2025). During Hurricane Helene (2024), TVA's strategies prevented approximately \$406M in potential damages (APPA 2024).

10.2 Data challenges

Since 1980, NOAA has provided a count of annual U.S. weather-related disasters that it estimates to have exceeded \$1 Billion (inflation adjusted), showing a substantial increase starting in 2008. NOAA and other government officials have cited the upward trend in the Billion Dollar Disaster series as evidence that climate change is making extreme weather worse (Pielke Jr., 2024). But over time, population and wealth have increased dramatically in the U.S., so when an extreme weather or climate event occurs, there is more damage even if there is no underlying trend in the frequency or intensity of extreme weather. Pielke Jr. (2024) demonstrates that losses per weather disaster as a proportion of GDP have decreased by about 80 percent since 1980, as shown in Figure 10.1. Pielke Jr. (2024) also argued that in addition to relying on opaque data sources and unreported adjustments, NOAA failed to normalize its Billion Dollar Disaster data series properly for changes in population exposure and wealth. In May 2025 NOAA announced it has withdrawn the Billion Dollar Disaster product from publication (Pielke Jr., 2025).

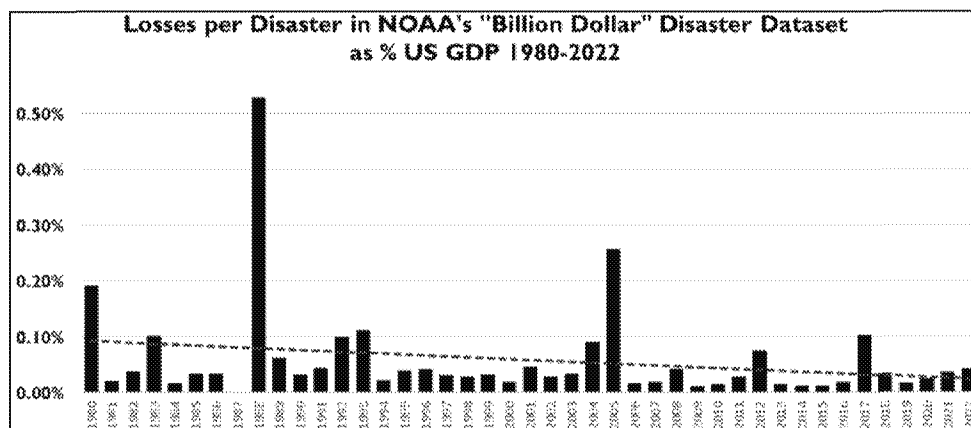


Figure 10.1: Losses per disaster as a % of Gross Domestic Product in NOAA's billion-dollar disaster dataset (the version downloaded in July 2023), 1980 to 2022. Source: (Pielke, Jr. 2024)

In summary, trends in losses from extreme weather and climate events are dominated by population increases and economic growth. Technological advances such as improved weather forecasting and early warning systems have substantially reduced losses from extreme weather events. Better building codes, flood defenses, and disaster response mechanisms have lowered economic losses relative to GDP. Finally, the U.S. economy's expansion has diluted the relative impact of disaster costs, as seen in the comparison of historical and modern GDP percentages.

10.3 Mortality from temperature extremes

10.3.1 Heat and cold risks

Changes in temperature extremes are among the most certain impacts expected in a warming world. It stands to reason that extreme heat events would likely become more frequent while extreme cold events would become less frequent. This pattern is evident in the historical period, though not in the continental U.S. (Chapter 6), and is expected to continue with further warming.

Mortality during heat extremes is typically caused by heat stroke and heat exhaustion, while mortality during cold extremes typically stems from hypothermia and heart strain. Global mortality is substantially greater for cold conditions than for hot conditions (Zhao et al 2021, Ritchie 2024). Unlike with heat-related mortality, cold-weather risks set in even at moderately cold conditions (Gasparini *et al.* 2015, Lee and Dessler 2023). The U.S. EPA (citing data from the Centers for Disease Control) reports that on average over 1999 to 2015 there were 2.2 deaths per million Americans for which cold was listed as the main underlying cause and an additional 2.4 deaths per million for which cold was listed as a contributing factor (EPA 2025). By contrast there were 1.3 deaths per million for which heat was the main cause and an additional 0.8 per million for which it was a contributor. By this metric cold accounts for approximately double the weather-related deaths as does heat.

Epidemiological methods that consider correlational evidence, not just death certificate reports, indicate the cold/warm ratio might be much higher. A 13-country study of 74 million deaths from 1985 to 2012 estimated that, on average, 7.7 percent of deaths were attributable to sub-optimal temperatures, of which 7.4 percent were attributable to cold and only 0.4 percent were attributable to heat (Gasparini *et al.* 2015). In other words, cold weather killed 18.5 times as many people as did hot weather.

Figure 10.2 shows the distribution of results from Gasparini *et al.* (2015) by country. For the United States the fraction of deaths attributable to temperature was 5.9 percent, of which 5.5 percent was due to cold, thus cold weather killed 14 times as many people as hot weather.

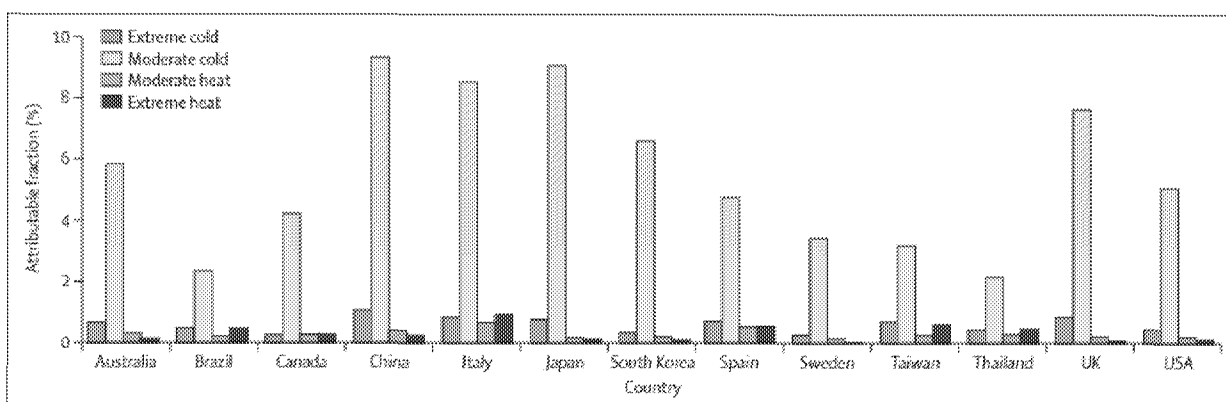


Figure 10.2. Mortality attributable to extreme and moderate cold and heat by country. Source: reproduced from Gasparini *et al.* (2015).

There is strong evidence that people adapt to weather risks. Lee and Dessler (2023) reported that 86 percent of temperature-related deaths across 40 cities in the U.S. were due to cold-related mortality, and that due to adaptation the relative risk of death declined in hot and cold cities alike as seasonal temperatures increased. Allen and Sheridan (2018) found that short, early-season cold events were 2 to 5 times deadlier than hot events, but the mortality risk of both cold and hot extremes drops to nearly zero if the events occur late in the season.

Davis *et al.* (2003) examined heat-related mortality in 28 U.S. cities from the 1960s to the end of the 1990s and found that heat-related mortality declined by three-quarters over the sample period. Bobb *et al.* (2014) examined mortality data for 106 million people in over 100 cities U.S. cities and found a 60 percent decline in average heat-related mortality over the period 1987–2005, from 51 per thousand deaths to 19. They furthermore found that the greatest drop was among seniors over the age of 75. In a study of 42 million

deaths in 211 U.S. cities from 1962 to 2006. Nordio *et al.* (2015) found a more than 90 percent decline in the risk of mortality from excess heat.

In the context of large declines in heat-related mortality, rising temperatures are associated with a net saving of lives since they reduce mortality from cold events. AR6 Working Group 2 Chapter 16.2.3.5 (O'Neill *et al.* 2022) acknowledges that heat-related mortality risk is declining over time:

Heat-attributable mortality fractions have declined over time in most countries owing to general improvements in health care systems, increasing prevalence of residential air conditioning, and behavioral changes. These factors, which determine the susceptibility of the population to heat, have predominated over the influence of temperature change.

Yet the IPCC misrepresents the overall situation in its AR6 Synthesis report. Section A.2.5 of that document states: “In all regions increases in extreme heat events have resulted in human mortality and morbidity (*very high confidence*).” But it is silent on the larger decline of deaths during extreme cold events.

The observed decline in U.S. heat-related mortality has been specifically attributed to adaptation. Wang *et al.* (2018) exploited the spatial variability of heatwave-related mortality across 209 U.S. cities from 1962 to 2006. While simple correlation appeared to imply an increase in mortality risk during heatwaves, accounting for adaptation to heatwave intensity caused the effect to fall to near zero and become statistically insignificant. They used the results of their epidemiological model to project heat-related mortality out to 2050 under four RCP warming scenarios (including RCP8.5) with and without adaptive behavior. Assuming people continue to adapt to the heatwave risks in the regions in which they live. Wang *et al.* (2018) project not only no increase in heat-related mortality, but an overall mortality *decrease* for the U.S. They conclude that

Ignoring adaptation would result in a substantial overestimate of future mortality related to heat waves... Accounting for adaptation, the overall heat-related mortality by 2050 would not change substantially over time compared to 2006.

10.3.2 Mortality risks and energy costs

A 2016 study of U.S. long term mortality risks associated with temperature variations (Barreca *et al.* 2016) showed that increases in mortality across the U.S. are associated with both cold weather and hot weather. But over time, the introduction of electricity and the adoption of central heating and air conditioning (AC) dramatically reduced both risks, especially those associated with hot weather. Prior to 1960 a day above 90°F (32°C) added 2.2 percent to the average mortality risk rate, but after 1960 the same weather added only 0.3 percent to mortality risk, an 85 percent reduction. Prior to 1960 temperatures below 39°F (4°C) added about 1 percent to mortality risk but after 1960 the same weather only added about half that amount. Adaptation through conventional household improvements dramatically reduced public vulnerability to weather extremes. The entire reduction in hot weather mortality was attributable to widespread adoption of indoor AC, which depended on the availability of reliable and affordable electricity.

The corollary of this finding is that the use of heating and cooling systems depends on energy being affordable. Doremus *et al.* (2022) showed that wealthy and poor households in the U.S. adjust their energy expenditures at similar rates in response to moderate temperature swings, but not in response to extreme temperature swings. When temperatures swing to very cold levels (< 5°C) energy spending in high-income

households rises by 1.2 percent but in low-income households by only 0.5 percent. On very hot days (>30°C) electricity spending in high-income households rises by 0.5 percent but does not change at all in low-income households. The latter result is observed even in subsamples where all households have AC. Cong et al. (2022) report similar findings for a sample of households in Arizona. The implication is that even with widespread adoption of home heating and cooling systems, inability to afford energy leaves low-income households exposed to weather extremes.

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11 CLIMATE CHANGE, THE ECONOMY, AND THE SOCIAL COST OF CARBON

Chapter summary

Economists have long considered climate a relatively unimportant factor in economic growth, a view echoed by the IPCC itself in AR5. Mainstream climate economics has recognized that CO₂-induced warming might have some negative economic effects, but they are too small to justify aggressive abatement policy and that trying to “stop” or cap global warming even at levels well above the Paris target would be worse than doing nothing. An influential study in 2012 suggested that global warming would harm growth in poor countries, but the finding has subsequently been found not to be robust. Studies that take full account of modeling uncertainties either find no evidence of a negative effect on global growth from CO₂ emissions or find poor countries as likely to benefit as rich countries.

Social Cost of Carbon (SCC) estimates are highly uncertain due to unknowns in future economic growth, socioeconomic pathways, discount rates, climate damages, and system responses. The SCC is not intrinsically informative as to the economic or societal impacts of climate change. It provides an index connecting large networks of assumptions about the climate and the economy to a dollar value. Some assumptions yield a high SCC and others yield a low or negative SCC (i.e. a social benefit of emissions). The evidence for or against the underlying assumptions needs to be established independently; the resulting SCC adds no additional information about the validity of those assumptions. Consideration of potential tipping points does not justify major revisions to SCC estimates.

11.1 Climate change and economic growth

11.1.1 Overview

It has long been noted that economies tend to do poorly in very cold and very hot regions, with the optimum somewhere in between (Nordhaus, 2006). This implies that warming will tend to be harmful in hot regions but beneficial in cool ones. Temperature-sensitive economic activity migrates, whenever possible, to where it is best suited, and society adapts to the local climate. Based in part on these

observations, in 1992 Thomas Schelling, then President of the American Economic Association, argued that any effects of climate change on U.S. economic activity would be small relative to the many other changes that would happen (Schelling 1992).

... Manufacturing rarely depends on climate, and where temperature and humidity used to make a difference, air conditioning has intervened. When Toyota chooses among Ohio, Alabama, and Southern California for locating an automobile assembly, geographical considerations are important, but not because of climate... Finance is little affected by climate; similarly for health care, or education, or broadcasting. Transportation can be affected, but improvements in all-weather landing and take-off in the last 30 years are greater than any differences that climate makes. If the average effect is a warming, iced waterways and snow removal may decline in importance. Construction is affected, mainly by cold, and if the average effect is in the direction of warming, construction may benefit slightly.

It is really agriculture that is affected. But even if agricultural productivity declined by a third over the next half century, the per capita GNP we might have achieved by 2050 we would achieve only in 2051. ...

I conclude that in the United States, and probably Japan, Western Europe, and other developed countries, the impact on economic output will be negligible and unlikely to be noticed.

Thirty years later virtually the identical point was made by the IPCC itself in the Fifth Assessment Report (Arent *et al.* 2014, emphasis added)

For most economic sectors, the impact of climate change will be small relative to the impacts of other drivers. Changes in population, age, income, technology, relative prices, lifestyle, regulation, governance, and many other aspects of socioeconomic development will have an impact on the supply and demand of economic goods and services that is large relative to the impact of climate change.

Evidence since AR5 does not change this assessment. Mohaddes *et al.* (2023) found that warm weather shocks have small negative effects on U.S. state-level output but not income, while cold weather shocks negatively affect both and the impacts are about four times larger, implying a shift to warmer conditions would, if anything, yield a net economic benefit for the U.S.

These statements are validated by experience. Since 1900, the average global surface temperature anomaly warmed 1.3°C, about as much as the IPCC predicts will occur in the next century under a moderate emissions scenario. But even as the globe warmed and the population quintupled, humanity prospered as never before. For example, global average lifespan went from thirty-two years to seventy-two years, economic activity per capita grew by a factor of seven, and the death rate from extreme weather events plummeted by a factor of fifty. Climate change damage projections typically refer to reductions in how much life will improve for humanity, they don't state that it will get worse in an absolute sense (O'Neill 2023).

While extreme weather events are costly, in all modern economies they are becoming steadily less and less important (Formetta and Feyen, 2019). Since 1990 weather-related disaster losses have declined as a proportion of global GDP (Pielke Jr , 2018, 2020) as have mortality risks (Formetta and Feyen, 2019). While economic weather-related insurance payouts are rising, this is fully explained by the growth in the size of the economy and the value of the insured asset stock. Past increases in episodes extreme weather

have not had a significant effect (positive or negative) on the market value of insurance firms (Hu and McKittrick, 2015). Nor have past extreme weather events had a significant effect on U.S. banks' performance (Blickle *et al.*, 2021); warming has even been shown to be beneficial for the finance and insurance sector (Mohaddes *et al.* 2023). Figure 11.1 below is illustrative. For these reasons, economists have long been reluctant to endorse attempts to “stop” climate change or even aggressively reduce GHG emissions because the costs would not be worth it. As one critic of the economics of climate policy put it (Storm 2017):

Mainstream climate economics takes global warming seriously, but perplexingly concludes that the optimal economic policy is to almost do nothing about it... The contrast is striking. While climate science is sending out loud-and-clear messages that fossil-fuel disinvestment must start now, letting go of coal and oil and diverting resources into renewable energy technology systems, to keep warming below the 2°C limit (IPCC 2014), mainstream climate economics claims that overly ambitious climate targets will unnecessarily hurt the economy and immediate de-carbonization is too expensive. Most climate economists thus recommend humanity to just wait-and-see.

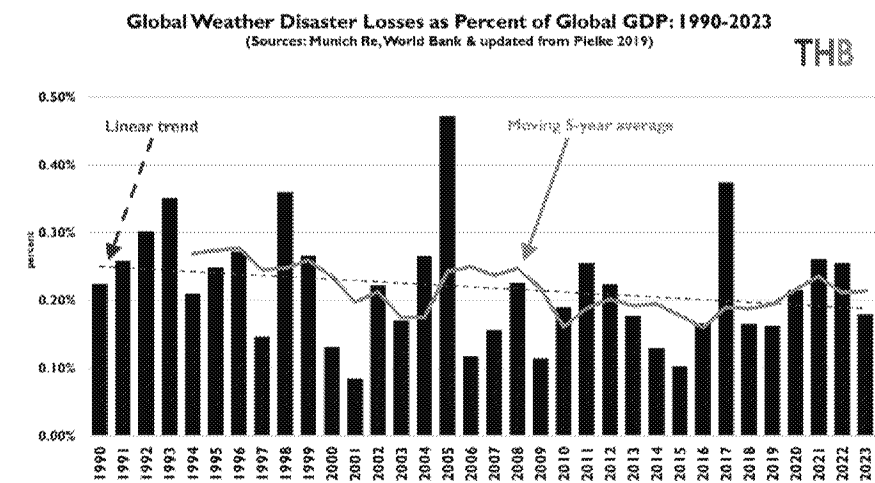


Figure 11.1 Global weather losses as a fraction of GDP. Source: Pielke Jr. (2023)

The mainstream economics position on the climate policy question is best represented by the findings over the past three decades from Integrated Assessment Models (IAMs) of climate change policy, for which Yale economist William Nordhaus was awarded the 2018 Nobel Memorial Prize in economics. IAMs combine economic, climate, and social data into a unified framework for simulating climate damages and evaluating the optimal policy response (Resources for the Future, 2025). Nordhaus' work has generally supported modest global climate policy, with aggressive measures largely deferred to later in the century. Nordhaus developed the so-called Dynamic Integrated Model of the Climate and Economy model, or DICE, in the early 1990s (Nordhaus, 1993) to study the interaction between climate change, climate policy, and global economic growth over long times. DICE assumes global emission control can be coordinated at no cost and asks what the optimal policy target should be. The climate component of the DICE model is based on simplified climatological modeling. The version of DICE at the time of Nordhaus' Nobel Prize award was described in Nordhaus (2018); it had a climate sensitivity parameter of 3.1°C for CO₂ doubling, consistent with the best estimate in IPCC reports (3.0°C).

The baseline (no policy) scenario incurs only \$0.4T in global abatement costs and leads to \$134.2T (trillion dollars) in global climate damages for a total cost of \$134.6T. The climate model component of DICE projects 4.1°C warming by 2100 relative to preindustrial temperatures. Note that this is a higher estimate of warming than many IPCC climate models.

The Optimal Policy scenario barely deviates from the business-as-usual path. It aims for +3.5°C warming, in other words we modestly scale back fossil fuel use and otherwise just live with almost all the warming. This suggests that most CO₂ emissions are less harmful than the policies that would be necessary to abate them. Trying to prevent warming causes costs quickly to outstrip the benefits. Pursuing a goal of capping warming at 2.5°C creates total costs of \$177.8T, which is \$43.2T worse than doing nothing at all. Nordhaus didn't evaluate trying to reach a Paris-type target of 1.5°C warming but it would be even more costly.

A subsequent edition of DICE includes higher assumed damages from warming which, not surprisingly leads to more aggressive policy recommendations, as explained in the section below on the Social Cost of Carbon.

11.1.2 Empirical analysis of climate change and economic growth

Many other studies have used econometric methods applied to historical data, instead of IAMs, to study the potential impact of climate change on economic growth. Dell *et al.* (2012, herein DJO12) was an influential study that used a multi-country panel of national-level data spanning 1950 to 2005, in which they matched climate and economic data by averaging temperature from the local grid cell level up to the national level using population weights. They found that warming yields an insignificant positive effect on income growth in rich countries but a significant negative effect in poor countries. Moore and Diaz (2015) modified the DICE model to take that finding into account and concluded that it implied a dramatically higher Social Cost of Carbon due to the compounding effects of income loss over time.

A large subsequent literature has debated the robustness of the DJO12 findings. Burke *et al.* (2015) analyzed a global panel with temperatures averaged up to the national level and found a negative effect on growth from warming in rich and poor countries alike when the national average temperature is above 13°C. Zhao *et al.* (2018) used the G-Econ data set of Nordhaus (2006), which breaks economic activity down to the grid cell level, replicated DJO12-type results on the same subset of countries as used in DJO12 but then showed that on the full global sample warming increases growth in rich countries and poor countries alike, though the positive effect in the latter group is confined to where local temperatures are below about 16°C. Greßer *et al.* (2021) developed a regional economic data set for 1,542 sub-national regions around the world between 2005 and 2015 and found temperature had no effect on growth. Yang *et al.* (2023) updated the DJO12 data set and applied an estimator robust to mixed sample frequencies, finding that while temperature shocks exerted a temporary effect on income levels, they did not have a significant lasting effect on growth rates.

Newell *et al.* (2021) noted that there is no underlying theory to guide econometric model specification in this literature. Taking into account the arbitrary choices of which explanatory variables to include, they identified over 800 possible model specifications. Using the Burke *et al.* (2015) data they used an estimation method that accounts for model uncertainty and found that the model form preferred by Burke *et al.* (2015), which implied negative effects of warming on growth even in rich countries, is explicitly excluded by the optimal model selection algorithm. Overall, they could not detect a temperature effect on GDP or GDP growth, and they estimated the 95 percent confidence interval for the impact on global growth as of 2100 even under the exaggerated RCP8.5 warming scenario spans –86 percent to +388 percent. In other words the net effect is likely positive but too uncertain to distinguish from zero.

Barker (2023) criticized the DJO12 assumption that countries can be grouped into fixed “poor” and “rich” categories based on their incomes many decades ago. He noted that many countries were once poor but became rich over time, and if this is considered the original temperature effects reported by Dell *et al.* became small and insignificant.

Berg *et al.* (2023) argued that countries shouldn’t be grouped into large Rich/Poor categories because they are too heterogeneous. They instead estimated country-specific temperature response coefficients then grouped countries with similar response coefficients into small panels. They separately estimated responses to global and idiosyncratic local temperature shocks to better identify the climate signal in weather data. Overall, they found countries experiencing negative effects of warming on growth outnumbered those experiencing positive effects, but only temporarily: eventually the effects reverse such that about twice as many countries experience a net positive growth effect. They also found global (as opposed to local) temperature changes are much more likely to benefit growth in poor countries than rich ones. In a simulation to 2100, even using the extreme RCP8.5 scenario, they computed the average global GDP loss would be only 1.9 percent compared to a scenario with no warming. That is, instead of the economy growing 400 percent it would grow 392%. The implication of Nordhaus’ earlier analysis is that trying to prevent the warming would lead to far less than 392 percent growth.

To summarize, economists consider climate a relatively unimportant factor in economic growth, a view echoed by the IPCC itself in the Fifth Assessment Report. Mainstream climate economics has recognized that CO₂-induced warming might have some negative economic effects but they are too small to justify aggressive abatement policy and trying to “stop” or cap global warming even at levels well above the Paris target would be worse than doing nothing. An influential study in 2012 suggested that global warming would harm growth in poor countries but the finding has subsequently been found not to be robust (Tol 2024). Studies that take full account of modeling uncertainties either find no evidence of a negative effect on global growth from CO₂ emissions or find poor countries as likely to benefit from it as rich countries.

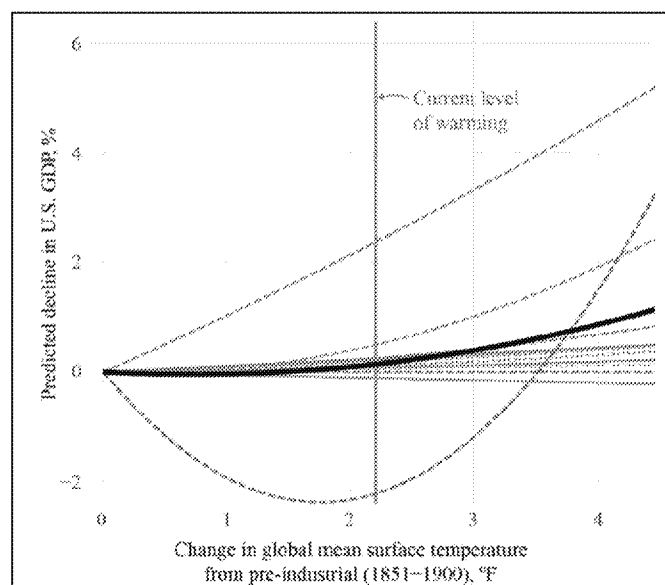


Figure 11.2: Decline in U.S. GDP per degree of warming. Source: CEA-OMB (2023)

The expectation that significant global warming would have a small impact on the U.S. economy was acknowledged quietly by the Biden Administration, even as the President was proclaiming a climate emergency. Figure 11.2, from a 2023 CEA-OMB report, shows the expected decrement in U.S. GDP as a

function of temperature rise. The colored lines show the results of a dozen peer-reviewed published studies while the solid black line is their average. The figure could be summarized as “a few percent impact for a few degrees of warming”. Given that the economy’s annual growth rate is expected to be 1-2 percent, the impact of a warming globe on the U.S. GDP is indeed negligible.

11.2 Models of the Social Cost of Carbon

The Social Cost of Carbon (SCC) is a tool for quantifying the economic impact of carbon dioxide emissions, helping policymakers weigh the costs and benefits of climate policies. It estimates the damage caused by emitting one additional ton of CO₂, expressed in dollars. More formally, the SCC is the discounted present value of the current and future marginal loss of economic welfare due to an additional ton of CO₂ entering the atmosphere.

11.2.1 *Estimating the SCC*

Although the literature refers to “estimates” of the SCC, it is not estimated in the way other economic statistics are estimated. For instance, data on market transactions including prices and quantities can be used to estimate the current inflation rate or the growth rate of per capita real Gross Domestic Product, and there are well-understood uncertainties associated with these quantities. But there are no market data available to measure many, if not most, of the marginal damages or benefits believed to be associated with CO₂ emissions, so these need to be imputed using economic models.

For example, an influential component of some SCC calculations is the perceived social cost associated with a changed risk of future mortality due to extreme weather. There is no market in which people can directly attach a price to that risk. At best economists can try to infer such values by looking at transactions in related markets such as real estate or insurance, but isolating the component of price changes attributable to atmospheric CO₂ levels is very difficult.

Economists use IAMs to compute the SCC. Two of the best-known are the Climate Framework for Uncertainty, Negotiation and Distribution (“FUND”, Tol 1997) and Nordhaus’ DICE. EPA (2023) introduced new ones for its recent work. IAMs embed a “damage function” or set of functions relating ambient temperature to local economic conditions. The assumptions embedded in the damage function will largely determine the resulting SCC. IAMs also assume a long-term discount rate or, as in DICE, compute the optimal internal discount rate as part of the solution.

One approach to developing a damage function is to begin with estimates of the costs (or benefits) of warming in specific sectors in countries around the world and aggregate up to a global amount. This was the approach used in FUND. An alternative approach is to develop a simple equation that penalizes global income according to a simple quadratic function of the average global temperature. This approach was used in DICE. In the case of the FUND model many hundreds of parameters had to be selected, whereas in DICE only three were needed and were originally chosen to assign a predetermined penalty (1.2%) to global output as of 3°C warming, with a quadratic term implying that damages grow with the square of the global average temperature anomaly. Barrage and Nordhaus (2024) recently changed the parameters to increase the penalty at 3°C to 1.6%, and added a discrete additional 1.0 percent GDP penalty at 3°C warming to account for “tipping points” (discrete large-scale environmental changes triggered by crossing a warming threshold) and a “judgmental adjustment” of 0.5 percent for excluded impacts at 3°C warming. Not surprisingly the newer version of DICE generates much higher SCC estimates than before.

The concepts of estimation and uncertainty do not readily apply to SCC calculations. No amount of data collection can change the fact that many components of the SCC are unknown and rely on judgment and opinion based on knowledge of the underlying literature on the physical effects of climate change. SCC calculations are thus best thought of as “if-then” statements: *if* the following assumptions hold, *then* the SCC is \$X per tonne. The list of “if” statements includes the premise that the world’s climate and economy work according to the representation in the IAM. One way this might fail to hold is in the timing of warming. Every IAM assumes a value of the Equilibrium Climate Sensitivity (ECS), which controls the amount of warming that results from CO₂ emissions and can be freely varied for the purpose of generating a distribution of SCC values associated with uncertainties over ECS. But as Roe and Baumann (2013) pointed out, time-to-equilibrium increases with the square of ECS, so an upward adjustment of the ECS parameter without an appropriate slowing down of the adjustment process can yield distorted present value damage estimates. In particular, the upper tail of warming associated with some commonly-used ECS distributions is physically impossible for even a thousand years into the future (Roe and Baumann 2013) yet in an IAM would be realized within a couple of centuries. Failure to align ECS with time-to-equilibrium will lead to an overestimate of the SCC value.

11.2.2 Variations in the SCC

Every level of the IAM calculation includes assumptions, some more influential than others. Key assumptions include the following.

- The discount rate: Climate damages accrue over a long time horizon and costs a century or two in the future need to be discounted to the present. The higher the discount rate the smaller the present value of future damages and vice versa. The discount rate represents the opportunity cost of spending money today rather than investing it and then having more to spend tomorrow. Some economists have argued for the use of very low discount rates in SCC calculations, resulting in policy recommendations that favor relatively large immediate investments in CO₂ emission reductions. The downside is that other investments could potentially earn a larger rate of return for society.
- Equilibrium Climate Sensitivity: IAMs have customarily employed a value of 3.0°C or 3.1°C following the IPCC’s guidance. The most recent data-driven ECS values tend to be lower than this (see Chapter 4). Dayaratna *et al.* (2017, 2020, 2023) have shown that use of lower empirically-derived ECS values dramatically lower the resulting SCC estimate, even when low discount rates are used.
- Damage function coefficients: IAMs assume CO₂ and warming cause net harms that increase exponentially with temperature. More recently, IAMs have also incorporated effects from assumed potential climate tipping points. The FUND model took limited but explicit account of CO₂ fertilization effects in agriculture. Since the coefficients were selected prior the publication of the current evidence of global greening and the magnitude of benefits to crops from elevated CO₂ (see Chapters 2 and 9) the growth effects are likely understated. The DICE model (and others) did not explicitly include any CO₂ fertilization benefit, except to the extent it was taken into account in the literature consulted when picking the damage function coefficients. The damage function in FUND contains a region in which low warming yields net benefits in many regions, a finding which is supported by econometric models of warming and growth (Berg *et al.* 2023) and econometric simulations of agricultural changes (McKittrick 2025). The DICE damage function, by construction, rules out net benefits at any warming level.
- Emission scenarios: IAMs generate SCC estimates that increase as the pre-existing concentration of CO₂ increases. Consequently the value of damages later in the century will be higher, depending on the assumed baseline emissions over the coming decades.

Some IAMs (such as DICE) include the cost to the economy of reducing CO₂ emissions in order to identify the SCC along an optimal growth path. If CO₂ emission reductions are assumed to be inexpensive, then the model will conclude that the optimal policy should aim for deeper emission cuts and vice versa.

It is informative to ascertain whether SCC results are invariant to changes in some assumptions. But when different assumptions result in higher or lower SCC values, the change in the SCC value does not provide *prima facie* evidence about the validity of the assumptions. For example, in 2023 the U.S. Environmental Protection Agency raised its preferred SCC value about 5-fold over the estimates it had issued ten years earlier. This is not because new data had been collected or better mathematical methods had been invented, but because new assumptions had been used, and the validity of those assumptions was a separate question. Tabulations on EPA (2023) p. 81 show that if assumptions similar to earlier analyses had been applied, the results would not have materially differed from before. One new assumption was that global agricultural damages were far higher than previously believed, based on an analysis in Moore *et al.* (2017). But as discussed in Chapter 9, McKittrick (2025) showed that Moore *et al.* (2017) had used a database in which half the CO₂ change observations were missing. When as many of those observations as possible were recovered from underlying sources and the analysis was rerun, the projected crop yield losses disappeared and turned into gains at all warming levels. Hence the portion of the EPA's SCC revision attributable to agricultural yield losses was unwarranted.

11.2.3 Evidence for low SCC

Chapter 2 reviewed evidence on climate change and greening, and Chapter 9 looked at climate change and U.S. agriculture. Evidence shows that CO₂ fertilization has a stronger beneficial effect on agriculture than was known at the time that IAMs like DICE and FUND were parameterized. Haverd *et al.* (2020) found the observed CO₂ fertilization rate has been almost double what had been predicted by crop models. Dayaratna *et al.* (2023) used the updated empirical ECS distribution estimate of Lewis (2022), which assimilates modern instrumental and paleoclimatic temperature records, and allowed for a 30 percent gain in the CO₂ fertilization benefit in the FUND model, and found that, even at a low discount rate of 2.5 percent, the median SCC as of 2050 is only \$18.67, with a 24 percent probability of the true value being negative. At a five percent discount rate the median SCC value is negative until the mid-2040s and at 2050 was only \$0.37 with a 49 percent probability of being negative. Thus, under reasonable assumptions a mainstream IAM using updated scientific inputs yields evidence consistent with the SCC not being significantly greater than zero.

It should also be noted that the SCC is focused on the social costs of CO₂ emissions from fossil fuel use. It is not intended to measure the private marginal benefits to consumers and society from the availability of fossil fuels. Public willingness to pay for fuels of all types indicates the value to society of reliable, abundant fossil energy. Tol (2017) estimates that the private benefit of carbon is large relative to the social cost. This can be illustrated by noting that the price of a gallon of gas indicates the marginal value to the consumer of the fuel. Suppose we assume a relatively high Social Cost of Carbon of, say, \$75 per tonne. Deflated by a MCPF⁴ value of 1.5 that would result in a carbon tax of \$50 per tonne, which equates to about 44 cents per gallon of gas (Lavelle, 2019). A pre-tax price of \$3.00 per gallon would imply the marginal social benefit of the fuel is nearly seven times the marginal social cost.

11.2.4 Tipping points

SCC calculations typically consider gradual impacts of a warming climate, such as slowly melting glaciers and increasing average temperature. A driver of potentially high values of the Social Cost of Carbon

⁴ Marginal Cost of Public Funds: the optimal carbon tax rate is the SCC divided by the MCPF (Sandmo 1977).

(SCC) is the introduction into models of discrete catastrophic outcomes associated with abrupt changes (Dietz *et al.*, 2021). They are often referred to as “tipping points.”

The term “tipping point” mingles two different physical concepts that pose different research challenges. Many physical systems are inherently stable unless acted upon with sufficient external energy. For example, an ice cap might remain intact over a wide range of temperatures but once the temperature crosses the 0°C threshold it melts. Such discontinuities are ubiquitous in nature and require an external force. Whether the force needs to be large relative to the size of the system depends on the underlying stability of the system.

A different type of tipping point is called a *bifurcation* and arises from the study of the internal dynamics of nonlinear systems (Crawford, 1991). Many systems have been observed to have more than one equilibrium point and can move between them with minimal or no external influence. For example, a weather system might have two different equilibrium states: one calm and one with a tornado. A transition from one to another can happen either with no external force or with a minuscule change, such as a flap of the proverbial butterfly’s wings (Shen *et al.* 2014). The term “tipping point” is sometimes used to mean a bifurcation of this type and implies instability inherent in the system itself, which is not necessarily dependent on outside forces. It depends, instead, on parameters of the system taking values that support the emergence of bifurcations (Crawford, 1991).

The two different concepts imply different research questions. Regarding the first we are interested in whether components of the climate system are susceptible to abrupt discontinuities in response to sufficiently large anthropogenic forcing. Regarding the second we are interested in whether the Earth’s climate system has inherent bifurcations that imply the possibility of abrupt transitions with or without external forcing.

Models have been developed that imply the second type is a possibility. Kypke *et al.* (2020) presented a simple climate model in which the GHG concentration is one of the parameters that controls the emergence of bifurcations of the Arctic climate to one with both cold and warm equilibria. If sufficiently high GHG forcing combined with a sufficiently high rate of ocean heat transport are imposed a bifurcation becomes possible. More generally GCMs have been observed to contain bifurcations and multiple equilibria (Brunetti and Ragon, 2023).

The possible existence of bifurcations in the Earth’s climate system implies abrupt transitions are possible, not just in response to large forcing but also to small perturbations. This places tipping points into the category of low-likelihood and potentially catastrophic events, such as large meteor strikes. A key question to ask is whether those kinds of tipping points can be predicted. Current research has not resolved that question (Dakos *et al.*, 2024) and indeed might not be able to since one implication of the “butterfly effect” is the existence of predictability boundaries of nonlinear systems (Palmer *et al.*, 2014). It is therefore not obvious how to incorporate such possibilities into SCC calculations. Small variations in assumptions will lead to arbitrarily large variations in the resulting SCC with no grounds for choosing among them. If such tipping points are possible the most appropriate stance for economic policy is to maximize resilience to any form of external catastrophe since it is unlikely we could predict it or prevent it from happening.

AR6 (WGI, Chapter 1) focuses mainly on the first type of tipping point, namely an abrupt change in response to external forcing. This is also the meaning associated with popular usage of the “tipping point” concept in discussions of climate change (see [[HYPERLINK "https://report-2023.global-tipping-points.org/what-is-a-tipping-point/"](https://report-2023.global-tipping-points.org/what-is-a-tipping-point/)] for example). As summarized by AR6, there is evidence of abrupt change in the paleoclimate record, and some of these events have been interpreted as tipping points. Some projections with Earth System Models for example have produced tipping points such as Amazon forest dieback in response to specified values of CO₂ concentration or temperature increases.

The alarm surrounding climate tipping points is reflected by *The Global Tipping Points Report* that was launched at COP28 on 6 December 2023 (Lenton *et al.*, 2023). It identifies more than 25 parts of the climate

system said to constitute tipping points. What gets classified as a climate “tipping point” is a moving target. The most common examples in the literature and assessment reports include: Greenland ice sheet disintegration, West Antarctic ice sheet disintegration, summertime disappearance of Arctic sea ice, Amazon rainforest dieback, coral reef dieoff, thawing of permafrost and methane hydrates, Atlantic Meridional Overturning Circulation collapse, boreal forest shift, West African monsoon shift, and Indian Monsoon shift.

All such tipping points require a certain amount of system instability to exhibit an abrupt transition in response to warming. For this reason, there seems to be very little to discriminate between a tipping event and natural climate variability. Natural climate variability has in the past produced shifts in the West African and Indian monsoons, Amazon forest and coral reef dieback, and disintegration of parts of the Greenland and West Antarctic ice sheets. These impacts can reverse on the decadal and century timescales associated with natural climate variability and ecosystem responses.

Some abrupt changes are potentially more consequential, including collapse of the West Antarctic ice sheet and collapse of the Atlantic Meridional Overturning Circulation (AMOC). AR6 WG1 Summary for Policy Makers states:

The Atlantic Meridional Overturning Circulation is very likely to weaken over the 21st century for all emission scenarios. While there is *high confidence* in the 21st century decline, there is only *low confidence* in the magnitude of the trend. There is *medium confidence* that there will not be an abrupt collapse before 2100. (C.3.4)

There is limited evidence for low-likelihood, high-impact outcomes (resulting from ice-sheet instability processes characterized by deep uncertainty and in some cases involving tipping points) that would strongly increase ice loss from the Antarctic Ice Sheet for centuries under high GHG emissions scenarios (e.g., SSP5-8.5). (B.5.2)

For SCC calculations, the research question implied by this type of tipping point is whether such events have been observed in the past in climate conditions similar to what we currently experience or will in the near future. AR6 finds little evidence for impending collapse of the Atlantic Meridional Overturning Circulation or the West Antarctic ice sheet. It finds there is no tipping point associated with Arctic Sea ice (AR6 Technical Summary p. 76)

Dietz *et al.* incorporated several potential tipping points (abrupt changes) into an SCC model and found they added about 25 percent to the estimate, mainly associated with thawing permafrost and release of methane hydrates. However, the IPCC considers this scenario *very unlikely* (AR6 Technical Summary p. 107).

In summary, there might be unknown bifurcation tipping points that are associated with natural climate processes, but this possibility does not translate into specific guidance on the SCC. There are potential abrupt change points in the climate system in response to warming, although the IPCC assigns low probabilities to most, including the largest ones. When these have been considered, the result is only a modest increase in the SCC value in the 21st century.

11.2.5 Are there alternatives?

It is increasingly being argued that the SCC is too variable to be useful for policymakers. Cambridge Econometrics (Thoung, 2017) stated it’s “time to kill it” due to uncertainties. The UK and EU no longer use SCC for policy appraisal, opting for “target-consistent” carbon pricing (UK Department for Energy Security and Net Zero 2022, Dunne 2017). However, the uncertainty of SCC estimates doesn’t mean that other regulatory instruments are inherently better or more efficient. Many emissions regulations (such as

electric vehicle mandates, renewable energy mandates, energy efficiency regulations and bans on certain types of home appliances) cost far more per tonne of abatement than any mainstream SCC estimate, which is sufficient to establish that they fail a cost-benefit test.

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12 GLOBAL CLIMATE IMPACTS OF U.S. EMISSIONS POLICIES

Chapter Summary

U.S. policy actions are expected to have undetectably small direct impacts on the global climate and any effects will emerge only with long delays.

12.1 The scale problem

The emissions rates and atmospheric concentrations of criteria air contaminants are closely connected because their lifetimes are short and their concentrations are small; when local emissions are reduced the local pollution concentration drops rapidly, usually within a few days. But the global average CO₂ concentration behaves very differently, since emissions mix globally and the global carbon cycle is vast and slow. Any change in local CO₂ emissions today will have only a very small global effect, and only with a long delay.

Following the emission of a pulse (release) of CO₂ into the atmosphere only about 40 ± 15 percent of the extra CO₂ will have been sequestered after twenty years. That fraction rises to 75 ± 10 percent after a thousand years, and the remainder will be gradually removed over the ensuing tens of thousands of years (Ciais *et al.*, 2013, pp. 472-473). Consequently, any reduction in U.S. emissions would only modestly slow, but not prevent, the rise of global CO₂ concentration. And even if global emissions were to stop tomorrow, it would take decades or centuries to see a meaningful reduction in the global CO₂ concentration and hence human influences on the climate.

Reducing the atmospheric stock of CO₂ would require emissions to fall below the natural sequestration rate, assuming the entire increase is anthropogenic. Since that rate has been averaging about 50 percent of emissions in recent decades, a reduction of global emissions by 50 percent would (at least temporarily) halt the rise in atmospheric CO₂. The 1997 Kyoto Protocol proposed to cap industrial nations' CO₂ emissions at a modest five percent below 1990 levels by the year 2012. Even though this policy was too difficult for most nations to implement, full compliance would not have substantially reduced atmospheric CO₂ levels. It would only have slightly slowed CO₂ growth, reaching the projected year 2100 level in 2105 instead (Wigley, 1998). Lomborg (2016) estimated that full compliance with the initial commitments in the Paris Accord would not stop warming, it would only prevent about 0.1C warming and delay hitting the baseline year 2100 temperature levels by about a decade.

Thus, in contrast with conventional air pollution control, even drastic local actions will have negligible local effects, and only with a long delay. The practice of referring to unilateral U.S. reductions as “combatting climate change” or “taking action on climate” on the assumption we can stop climate change therefore reflects a profound misunderstanding of the scale of the issue.

12.2 Case study: U.S. motor vehicle emissions

The scale problem can be illustrated with reference to U.S. motor vehicles. The EPA's 2009 Endangerment Finding focused on CO₂ emissions from cars and light-duty trucks in the U.S. because Section 202(a) of Clean Air Act mandates the EPA to set emissions standards for motor vehicles if pollutants are found to endanger public health or welfare. The 2009 Endangerment Finding therefore obligated the EPA to regulate emissions from new motor vehicles, ostensibly to reduce or eliminate climate-related harms to the U.S. public.

Two questions that naturally arise are: (1) How large a reduction in CO₂ emissions would result from such regulation? and (2) What would be the climate impact of such regulation?

The first question can be addressed by comparing U.S. vehicle-based CO₂ emissions to the global total. The second question can be addressed by using the fact that the reduction in global warming would be, according to the models relied upon by the EPA, proportional to the reduction in global emissions, keeping in mind that the change in the CO₂ content of the atmosphere in any given year is the result of total global CO₂ emissions, not just U.S. emissions.

In 2022, the emissions from U.S. cars and light duty trucks totaled 1.05 billion metric tons of carbon dioxide (GtCO₂, EPA 2024). Meanwhile global CO₂ emissions from energy use totaled 34.6 GtCO₂ (Energy Institute 2024). Hence U.S. cars and light trucks account for only 3.0 percent of global energy-related CO₂ emissions. To a first approximation we can say that even eliminating *all* U.S. vehicle-based emissions would retard the accumulation of CO₂ in the atmosphere by a year or two over a century.

It would also reduce the overall warming trend by at most about 3 percent. For the period 1979-2023, which has the most extensive global coverage of a variety of weather data types, warming trends are determined to a precision of about ± 15 percent, so the impact of reducing the rate of global warming by eliminating U.S. vehicle CO₂ emissions would be far below the limits of measurability. Given that global-average temperature is the most direct climate change metric, impacts on any secondary climate metrics (e.g. severe weather, floods, drought, etc.) from reducing U.S. vehicle CO₂ emissions would be even less measurable.

Consequently, in contrast to the case of local air contaminants like particulates and ozone, even the most aggressive regulatory actions on GHG emissions from U.S. vehicles cannot be expected to remediate alleged climate dangers to the U.S. public on any measurable scale.

12.3 Concluding thoughts

This report supports a more nuanced and evidence-based approach for informing climate policy that explicitly acknowledges uncertainties. The risks and benefits of a climate changing under both natural and human influences must be weighed against the costs, efficacy, and collateral impacts of any “climate action”, considering the nation’s need for reliable and affordable energy with minimal local pollution. Beyond continuing precise, un-interrupted observations of the global climate system, it will be important to make realistic assumptions about future emissions, re-evaluate climate models to address biases and uncertainties, and clearly acknowledge the limitations of extreme event attribution studies. An approach that acknowledges both the potential risks and benefits of CO₂, rather than relying on flawed models and extreme scenarios, is essential for informed and effective decision-making.

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GLOSSARY

2XCO₂: Doubling of the atmospheric carbon dioxide (CO₂), a commonly used baseline for addressing climate change. In 2025, atmospheric CO₂ concentrations were approximately 50 percent of the way to 2XCO₂ compared to 1800.

ABATEMENT: In the context of environmental policy, the reduction of emissions.

ACE (Accumulated Cyclone Energy): A statistical measure of a tropical cyclone's energy accumulated over its lifetime, calculated from a summation of the square of maximum sustained wind speeds.

ACIDIFICATION: A commonly used term for the reduction in ocean pH from more alkaline to less alkaline values.

ACRIM (Active Cavity Radiometer Irradiance Monitor) GAP: the gap in total solar irradiance data between 1989 and 1991 caused by a delay in the launch of a satellite-borne monitor in time to overlap and intercalibrate with the prior system.

AEROSOLS: Tiny solid or liquid particles suspended in the air.

ALKALINE: Having a chemical pH greater than 7.0. The “opposite” of acidic, which means a pH less than 7.0.

AMBIENT: Relating to the immediate surroundings of something.

AMO: Atlantic Multidecadal Oscillation, a 60-year cycle of sea surface temperatures in the North Atlantic.

ANTHROPOGENIC: Created by humans.

AR (Assessment Reports): Periodic assessments published by the Intergovernmental Panel on Climate Change evaluating current knowledge of the climate system, global change, and associated issues. An AR now consists of reports from three working groups, the first (WG1) concerned with physical science and the others concerned with societal impacts and mitigation strategies. The most recent was AR6.

ATTRIBUTION: Assertion of a causal relation, chiefly from anthropogenic GHG emissions to observed climate patterns.

AUTOCORRELATION: The phenomenon in which the current value of a random time series variable is correlated with its value in a previous time period.

BIOMASS: The total quantity or weight of organic matter in a given area or volume.

CARBON: The 6th element in the periodic table. Popularly used as a shorthand term for carbon dioxide.

CMIP (Coupled Model Intercomparison Project): The Intergovernmental Panel on Climate Change's organized effort to compare many different global ocean-atmosphere climate models when they are all forced with the same emissions scenario(s). The two most recent were the CMIP5 and CMIP6 for, respectively, AR5 and AR6.

CLIMATE CHANGE: A change in climate due to human influences, as opposed to climate variability, which is natural in origin.

CONUS: The contiguous 48 U.S. states.

DICE (Dynamic Integrated model of the Climate and Economy): An Integrated Assessment Model developed by William Nordhaus in 1993.

DISCOUNT RATE: The interest rate used to convert future costs and benefits into their present value. Varies inversely with present value so a high discount rate reduces the present value of future costs and benefits.

DRY ICE: Frozen carbon dioxide.

ECONOMETRICS: The branch of economics concerned with the use of statistical methods (chiefly multiple regression) to analyse economic systems.

ECS (Equilibrium Climate Sensitivity): The total amount of eventual global-average surface warming in response to a hypothetical doubling of the atmospheric CO₂ concentration compared to pre-Industrial levels.

EL NIÑO: The warm phase of the El Niño – Southern Oscillation, involving weaker upwelling of cold water off the South American coast and weaker Pacific trade winds.

EMISSIONS SCENARIO: An assumed scenario of future greenhouse gas emissions (or resulting atmospheric concentrations) based upon assumptions regarding global economic activity, the prevalence of fossil fuel use, and (sometimes) global carbon cycle model estimates of the rate of CO₂ uptake by land and ocean.

ENDANGERMENT FINDING (EF): The 2009 finding by the Administrator of the Environmental Protection Agency that emissions of well-mixed greenhouse gases (primarily CO₂) endanger human health and welfare.

ENSO: El Niño – Southern Oscillation, a prominent natural climate fluctuation involving year-to-year variations in the upwelling of cold water off the equatorial coast of South America and associated strengthening or weakening of trade winds across the Pacific Ocean.

FACE: (Free-Air CO₂ Enrichment) Large-scale open-air studies that expose plants to elevated CO₂ levels, simulating future climate scenarios.

FOSSIL FUEL: A natural hydrocarbon fuel such as coal, oil, or natural gas formed in the geological past from the remains of living organisms.

FUND (Framework for Uncertainty, Negotiation, and Distribution): An Integrated Assessment Model developed by Richard Tol in 1997.

GBR (Great Barrier Reef): The world's largest coral reef ecosystem located off the northeast coast of Australia.

GDP (Gross Domestic Product): Total market value of all the final goods and services produced within a country's border, usually over one year.

GHCN (Global Historical Climatology Network): A time-varying number of globally distributed weather stations making hourly, daily, or monthly measurements of precipitation, temperature and sometimes snowfall and snow depth.

GLOBAL GREENING: The satellite-observed increase in greenness in most vegetated land areas observed since the early 1980s.

GNP (Gross National Product): Total market value of all the final goods and services produced by the citizens of a country, including exports and imports, usually over one year.

GDP (Gross National Product): Total market value of all the domestically-produced final goods and services produced by the citizens of a country, usually over one year.

GREENHOUSE EFFECT (GHE): The tendency for any planetary atmosphere containing greenhouse gases to be warmer in the lowest layers than if those gases did not exist.

GREENHOUSE GAS (GHG): An atmospheric gas that absorbs and emits infrared radiation, especially water vapor, CO₂, and methane.

GROSS PRIMARY PRODUCTION: The total chemical energy used by plants during photosynthesis over a specified period of time. Net Primary Production is the residual after using the energy required for respiration and represents the contribution to biomass growth.

HURST PHENOMENON: See long term persistence.

HYDROLOGY: The study of the movement of water, especially on land and in the atmosphere.

IAM (Integrated Assessment Model): A computerized tool that combines economics, climate science, and social sciences to provide quantitative relationships between human and Earth systems, helping to inform policy decisions.

IEA (International Energy Agency): A Paris-based intergovernmental organization, established in 1974, that provides policy recommendations, analysis and data on the global energy sector.

IPCC (Intergovernmental Panel on Climate Change): A panel of international experts and governmental representatives established by the United Nations and World Meteorological Organization in 1988 to provide governments at all levels with scientific information that they can use to develop climate policies.

IR (Infrared): Heat radiation that all solid objects and greenhouse gases emit by virtue of their temperature.

LA NIÑA: The cool phase of ENSO, involving stronger upwelling of cold water off the South American coast and stronger Pacific trade winds.

MAIZE: Corn.

MODEL: A computer code or collection of codes which quantify our knowledge of various processes and the interactions between those processes.

MORTALITY: Death or number of deaths, usually expressed for a specific population and period.

NCA (National Climate Assessment): A periodic report mandated by the Global Change Research Act of 1990 summarizing the impact of climate change on the United States. There have been five NCA reports: NCA1 (2000), NCA2 (2009), NCA3 (2014), NCA4 (2017-18), and NCA5 (2023).

OECD (Organization for Economic Cooperation and Development): A group of 38 member countries that have a democratic system of government and free economic systems.

PALEOCLIMATE: Climate of the distant past, before modern weather instruments were widely used (typically pre-1850), necessitating the use of “proxy” measurements such as tree rings, ice core, pollen records, etc.

PDI (Power Dissipation Index): A statistical estimate of the accumulated destructive potential of a tropical cyclone over time, calculated as a summation of the cube of the maximum sustained wind speeds.

PDO: Pacific Decadal Oscillation, a recurring ocean-atmosphere interaction centered over the Pacific basin associated with predominantly warm or cool patterns that influence the global climate.

PHOTOSYNTHESIS: The process by which plants grow, requiring carbon dioxide, water, and light at sufficiently warm temperatures (10 to 35 deg. C for optimal function).

PPM: Parts per million, a measure of the concentration of a substance in (for example) the atmosphere

PRE-INDUSTRIAL: The near-term historical period before which human greenhouse gas emissions were considered significant, usually assumed to be the mid-1700s.

RADIATIVE FEEDBACK: A change in the balance between absorbed solar radiation and emitted infrared (IR) radiation, referenced to the top of the atmosphere, caused by changes in surface temperature.

RADIATIVE FORCING: The change in the balance between absorbed solar radiation and emitted infrared (IR) radiation, referenced to the top of the atmosphere, caused by changes in greenhouse gases, anthropogenic aerosols, volcanoes, etc.

RCP (Representative Concentration Pathway) SCENARIOS: Different scenarios for future greenhouse gas emissions and their impact on the atmosphere, introduced in the IPCC's Fifth Assessment Report (AR5). RCP scenarios have a number label (e.g. RCP6.0) indicating how much radiative forcing they assume in 2100 relative to pre-Industrial times, in Watts per square meter. RCP4.5 and RCP6.0 are intermediate scenarios, whereas RCP8.5 is an extreme scenario with very large future emissions of greenhouse gases.

REGRESSION: A statistical method for selecting the coefficients of a linear equation to explain the behaviour of a dependent variable in terms of variations in one or more explanatory variables.

RICARDIAN ANALYSIS: A method used in environmental economics named after David Ricardo to estimate the economic impacts of climate change, particularly on agriculture, by making use of the hypothesis that expected changes in the return to future agricultural activity will be capitalized into current land values.

SCC (Social Cost of Carbon): An estimate of the net economic damages, measured in present-value dollars, caused by emitting one additional ton of CO₂ into the atmosphere.

SSP (Shared Socioeconomic Pathway) SCENARIOS: Different scenarios for future greenhouse gas emissions and their impact on the atmosphere, introduced in the IPCC's Sixth Assessment Report (AR6). Some of the SSP scenarios are analogous to the older RCP scenarios, for purposes of continuity and comparison to earlier climate model assessments.

TC: Tropical cyclone.

TMAX: The daily maximum surface air temperature.

TMIN: The daily minimum surface air temperature.

TOA: Top of atmosphere.

TONNE: Metric ton.

TOXICOLOGY: The study of the adverse effects of chemical substances on living organisms.

TREND: A coefficient, usually estimated using linear regression, representing the slope of a line of best fit through a time series of data, indicating any tendency for the series mean to move up or down over time.

TSI (Total Solar Irradiance): The measure of the total amount of incident solar radiation per unit area, including all wavelengths of electromagnetic radiation, that reaches the Earth's atmosphere. Typically shown in units of watts per square meter (W/m²).

UHI (Urban Heat Island): The tendency for inhabited areas to be warmer than their rural surroundings, due to the replacement of natural land and vegetation with roads, parking lots, buildings, and waste heat sources.

USHCN: The U.S. Historical Climatology Network, consisting of quality-controlled temperature and precipitation data from 1,218 surface weather stations.

METADATA FOR FIGURES AND TABLES

Figure 2.1 Screen shot from Figure 3 of Zhu *et al.* 2016 [HYPERLINK "<https://www.nature.com/articles/nclimate3004>"]

Figure 2.2 Screen shot from Gerhart and Ward (2010) Figure 2 [HYPERLINK "<https://nph.onlinelibrary.wiley.com/doi/pdf/10.1111/j.1469-8137.2010.03441.x>"]

Figure 2.3 Screenshot from Copernicus Marine Services [HYPERLINK "https://data.marine.copernicus.eu/product/GLOBAL_OMI_HEALTH_carbon_ph_area_averaged/description"]

Figure 2.4 Screen shot from AIMS (2023) Figures 2-4 [HYPERLINK "https://www.aims.gov.au/sites/default/files/2022-08/AIMS_LTMP_Report_on%20GBR_coral_status_2021_2022_040822F3.pdf"]

Figure 3.1.1 Screen shot of AR6 WG1 Ch2 Fig 10

Figure 3.1.2 Screen shot of AR6 WG1 Ch7 Fig 7-6

Figure 3.1.3 Author created figure. CO2 data source: [HYPERLINK "<https://gml.noaa.gov/ccgg/trends/index.html>"]

Figure 3.2.1 Screen shot from Hausfather *et al.* (2019) Figure S4 [HYPERLINK "<https://doi.org/10.1029/2019GL085378>"]

Figure 3.2.2 Author created figure. Data source: Freidlingstein *et al.* (2024) [HYPERLINK "<https://essd.copernicus.org/preprints/essd-2024-519>"]
Data source: [HYPERLINK "<https://globalcarbonbudget.org/download/1442/>"]

Figure 3.2.3 Author-created using OLS trends, Data source: [HYPERLINK "<https://globalcarbonbudget.org/download/1442/>"]

Figure 3.2.4 Author-created using OLS trends, Data source: [HYPERLINK "<https://globalcarbonbudget.org/download/1442/>"]

Figure 4.1 Screen shot from Scafetta (2021) Figure 1 [HYPERLINK "<https://doi.org/10.3390/cli9110161>"]

Figure 5.1 Data source: [HYPERLINK "<https://climexp.knmi.nl/start.cgi>"]. Author-created figure. One ensemble member per model, yearly average temperatures are computed from the global monthly raw temperatures; ‘range’ is the warmest model temperature minus the coolest model temperature in each year; standard deviation computed across all 33 models’ raw temperatures in each year.

Figure 5.2 Screen shot from Scafetta (2023) Figure 2 [HYPERLINK "<https://doi.org/10.1007/s00382-022-06493-w>"]

Figure 5.3 Author created figure. Data source: [HYPERLINK "<https://climexp.knmi.nl/start.cgi>"]. One SSP370 ensemble member per model. Linear OLS trends computed from global monthly temperature anomalies relative to 1981-2010.

Figure 5.4 Author created figure, based on Figure 3 of McKittrick and Christy (2020) [[HYPERLINK "https://doi.org/10.1029/2020EA001281"](https://doi.org/10.1029/2020EA001281)]. Updated using same methods and data extended from 2014 to 2024, see McKittrick and Christy (2025) in chapter references.

Figure 5.5 Screen shot of IPCC AR5 Figure 10.SM1, with author annotations

Figure 5.6 Author created figure following Christy and McNider (2017) [[HYPERLINK "https://doi.org/10.1007/s13143-017-0070-z"](https://doi.org/10.1007/s13143-017-0070-z)] with updated data available at [[HYPERLINK "https://www.nsstc.uah.edu/data/cmip6/ERA5_REANALYSES/"](https://www.nsstc.uah.edu/data/cmip6/ERA5_REANALYSES/)] and [[HYPERLINK "https://www.nsstc.uah.edu/data/cmip6/JRA3Q_REANALYSES/"](https://www.nsstc.uah.edu/data/cmip6/JRA3Q_REANALYSES/)] compiled here [[HYPERLINK "https://www.nsstc.uah.edu/data/cmip6/CWG25_Fig_5.7_Table_250627.xlsx"](https://www.nsstc.uah.edu/data/cmip6/CWG25_Fig_5.7_Table_250627.xlsx)]

Figure 5.7 Screen shot from [[HYPERLINK "https://climate.rutgers.edu/snowcover/chart_seasonal.php?ui_set=nhl&ui_season=1"](https://climate.rutgers.edu/snowcover/chart_seasonal.php?ui_set=nhl&ui_season=1)] (accessed May 27, 2025)

Figure 5.8 Screen shot from Rugenstein and Hakuba (2023) Figure 1 [[HYPERLINK "https://doi.org/10.1029/2022GL101802"](https://doi.org/10.1029/2022GL101802)]

Figure 5.9 Author created figure. Data source: [[HYPERLINK "https://climexp.knmi.nl/start.cgi"](https://climexp.knmi.nl/start.cgi)]. OLS linear trends. One SSP370 ensemble member per model, averaged surface air temperatures for June through August over 31N to 49N latitude, 84W to 102W longitude. Observed temperature data for same months from NOAA/NCEI at <https://www.ncei.noaa.gov/access/monitoring/climate-at-a-glance/statewide/time-series> for the 12 Corn Belt states averaged together.

Figure 6.1.1 Author replot of Figure 10 from Koutsoyiannis (2013) [[HYPERLINK "http://dx.doi.org/10.1080/02626667.2013.804626"](http://dx.doi.org/10.1080/02626667.2013.804626)]

Figure 6.2.1 Screen shot of figure from Maue (2025) [[HYPERLINK "https://climatlas.com/tropical/"](https://climatlas.com/tropical/)]

Figure 6.2.2 Author created figure. Data from National Hurricane Center (2024) [[HYPERLINK "https://www.nhc.noaa.gov/climo/"](https://www.nhc.noaa.gov/climo/)]

Figure 6.2.3 Author created figure. Data from NOAA HRD [[HYPERLINK "https://www.aoml.noaa.gov/hrd/hurdat/All_U.S._Hurricanes.html"](https://www.aoml.noaa.gov/hrd/hurdat/All_U.S._Hurricanes.html) \t "_new"]

Table 6.2.1 Author created table. Data from NOAA HRD [[HYPERLINK "https://www.aoml.noaa.gov/hrd/hurdat/most_intense.html"](https://www.aoml.noaa.gov/hrd/hurdat/most_intense.html) \t "_new"]

Figure 6.3.1 Screen shot from NCA4 Figure 6.4

Figure 6.3.2 Author created figure. Data source: [[HYPERLINK "https://www.nsstc.uah.edu/data/ushcn_jrc/"](https://www.nsstc.uah.edu/data/ushcn_jrc/)]

Figure 6.3.3 Author created figure. Data source: [[HYPERLINK "https://www.nsstc.uah.edu/data/ushcn_jrc/"](https://www.nsstc.uah.edu/data/ushcn_jrc/)]

Figure 6.3.4 Author created figure. Data source: [[HYPERLINK "https://www.nsstc.uah.edu/data/ushcn_jrc/"](https://www.nsstc.uah.edu/data/ushcn_jrc/)]

Figure 6.3.5 Author created figure Data source: [HYPERLINK "https://www.nsstc.uah.edu/data/ushcn_jrc/"]

Figure 6.3.6 Author created figure. Data source: [HYPERLINK "https://www.nsstc.uah.edu/data/ushcn_jrc/"]

Figure 6.3.7 Screen shot from [HYPERLINK "https://www.globalchange.gov/indicators/heat-waves"]

Box tables. Author created table. Data from McKittrick and Christy (2025)

Figure 6.4.1 Author created figure. Data from McKittrick and Christy (2025)

Figure 6.4.2 Author created figure. Data from McKittrick and Christy (2025)

Figure 6.4.3 Author created figure. Data from McKittrick and Christy (2025)

Figure 6.4.4 Author created figure. Data from McKittrick and Christy (2025)

Figure 6.4.5 Author created figure. Data from McKittrick and Christy (2025)

Figure 6.5.1 Author created figure. Data source: [HYPERLINK "https://www.spc.noaa.gov/wcm/data/1950-2024_actual_tornadoes.csv"]

Figure 6.7.1 Author created figure. Data source: [HYPERLINK "https://www.ncei.noaa.gov/access/monitoring/uspa/wet-dry/0"] accessed June 16, 2025. OLS trend line added.

Figure 6.8.1 Screen shot from Lizundia-Loiola *et al.* (2021) Figure 12.

Figure 6.8.2 Screenshot from Marlon *et al.* (2012) Figure 2 panel C.
[HYPERLINK "https://www.pnas.org/doi/pdf/10.1073/pnas.1112839109"]

Figure 6.8.3 Author created Figure. Data 1926 to 2016: [HYPERLINK "https://web.archive.org/web/20200212033452/https://www.nifc.gov/fireInfo/fireInfo_stats_totalFires.html"]. Post-2017 [HYPERLINK "https://www.nifc.gov/fire-information/statistics/wildfires"] Accessed June 16, 2025.

Figure 7.1 Screenshot from [HYPERLINK "https://tidesandcurrents.noaa.gov/sltrends/"]

Table 7.1 Author created table. From Section 6.2 of
[HYPERLINK "https://judithcurry.com/wp-content/uploads/2018/11/special-report-sea-level-rise-3.pdf"]

Figure 7.2 Screenshot from [HYPERLINK "https://tidesandcurrents.noaa.gov/sltrends/sltrends_station.shtml?id=9414290"] (downloaded 4/22/25)

Figure 7.3 Screenshot from
[HYPERLINK "https://tidesandcurrents.noaa.gov/sltrends/sltrends_station.shtml?id=8771450"] (downloaded 4/22/2025)

Figure 7.4 Screenshot from

[HYPERLINK "https://tidesandcurrents.noaa.gov/sltrends/sltrends_station.shtml?id=8761724"]
(downloaded 4/22/25)

Figure 7.5 Screenshot from

[HYPERLINK "https://tidesandcurrents.noaa.gov/sltrends/sltrends_station.shtml?id=8518750"]
(downloaded 4/22/25)

Figure 7.6 Author created figure. Data from [HYPERLINK

"https://tidesandcurrents.noaa.gov/sltrends/sltrends_station.shtml?id=8518750"]

Figure 8.1 Screenshot from

[HYPERLINK "<https://crudata.uea.ac.uk/cru/data/temperature/HadCRUT5.0Analysis.pdf>"]

Figure 8.2 Screenshot from Hansen and Karecha (2025) Figure 1

[HYPERLINK "<https://www.columbia.edu/~jeh1/mailings/2025/CloudFeedback.13May2025.pdf>"]

Table 8.1 Screenshot of column 1 of Table 12.12 IPCC AR6 Working Group I report

Figure 9.1 Screenshot From Taylor and Schlenker (2021) Figure 1

[HYPERLINK "<https://www.nber.org/papers/w29320>"]

Figure 9.2 Screenshot from McKittrick (2025) Figure 1

[HYPERLINK "<https://www.nature.com/articles/s41598-025-90254-2>"]

Figure 10.1 Screenshot from Pielke (2024) [HYPERLINK "<https://www.nature.com/articles/s44304-024-00011-0>"] Figure 3.

Figure 10.2 Screenshot from Gasparini *et al.* (2015) Figure 2

[HYPERLINK "[http://dx.doi.org/10.1016/S0140-6736\(14\)62114-0](http://dx.doi.org/10.1016/S0140-6736(14)62114-0)"]

Figure 11.1 Screen shot from Pielke Jr (2023)

[HYPERLINK "<https://rogerpielkejr.substack.com/p/global-disaster-losses1990-2023>"]

Figure 11.2 Screenshot from CEA-OMB (2023) [HYPERLINK

"<https://bidenwhitehouse.archives.gov/wp-content/uploads/2023/03/CEA-OMB-White-Paper.pdf>"]
Figure 1.

ABOUT THE AUTHORS

John Christy, Ph.D. is the Distinguished Professor of Atmospheric and Earth Sciences and Alabama's State Climatologist at the University of Alabama in Huntsville. He has a B.A. (Mathematics) from CSU Fresno and an M.S. and Ph.D. (Atmospheric Sciences) from the University of Illinois in Urbana-Champaign. He is a Fellow of the American Meteorological Society and recipient of The NASA Medal for Exceptional Scientific Achievement for constructing, with Dr. Roy Spencer, the first global atmospheric temperature dataset from polar orbiting satellites. He served as Lead Author of the 3rd IPCC Assessment Report (2001), he coauthored the National Academy of Science expert report on Surface Temperature Reconstructions (2006), and he has served on the NOAA Climate Change Science Program, NASA's Earth Science Subcommittee, the National Research Council Space Studies Board and the EPA Science Advisory Board. In addition, Dr. Christy has given testimony in 20 Congressional Hearings.

Judith Curry, Ph.D. is Professor Emerita at the Georgia Institute of Technology, where she served as Chair of Earth and Atmospheric Sciences for 13 years. She is President and co-founder of Climate Forecast Applications Network (CFAN). Curry has a PhD in Geophysical Sciences from the University of Chicago. She has authored or coauthored 192 peer-reviewed papers in atmospheric and climate sciences, two textbooks, and most recently the book *Climate Uncertainty and Risk*. Curry is a Fellow of the American Meteorological Society, the American Association for the Advancement of Science, American Geophysical Union, and a Member of the American Academy of Sciences and Letters. She has served on several Science Steering Committees of the World Climate Research Program and also the DOE Biological & Environmental Research Advisory Committee, Earth Science Subcommittee of the NASA Advisory Council, and the NRC Space Studies Board and Climate Research Committee. Dr. Curry has given testimony in 13 Congressional Hearings..

The Honorable Steven E. Koonin, Ph.D. is the Edward Teller Senior Fellow at Stanford's Hoover Institution. He has served as Under Secretary for Science at the Department of Energy (2009-2011), as Chief Scientist for British Petroleum (2004-2009), and as a professor at Caltech (1975 – 2004, the last nine years as Vice President and Provost). Koonin is a member of the National Academy of Sciences and the JASON group of government advisors, as well as a Governor of Lawrence Livermore National Laboratory, having served in similar capacities for the Los Alamos, Sandia, Brookhaven, and Argonne National Laboratories. Other prior roles include a University Professor at New York University, a non-resident Senior Fellow at the American Enterprise Institute, and a Trustee of the Institute for Defense Analyses. Koonin has a BS in Physics from Caltech and a PhD in Theoretical Physics from Massachusetts Institute of Technology. He authored the 2021 bestseller *Unsettled: What Climate Science Tells Us, What It Doesn't, and Why It Matters*, the 1985 textbook *Computational Physics*, and some 200 peer-reviewed papers in physics and astrophysics, scientific computation, energy technology and policy, and climate science.

Ross McKittrick, Ph.D., is a Professor of Environmental Economics at the University of Guelph in Ontario, Canada. He holds a BA in economics from Queen's University, and an MA and Ph.D. in economics from the University of British Columbia. He has published widely in both economics and physical science journals on topics related to climate change, pollution and public policy. His book *Economic Analysis of Environmental Policy* was published by the University of Toronto Press in 2010. His background in applied statistics has led him to author or coauthor peer-reviewed publications across a wide range of topics in the physical sciences including paleoclimate reconstruction, malaria transmission, surface temperature measurement, climate attribution methodology and climate model evaluation. Professor McKittrick has made many invited academic presentations around the world, was an invited contributor to the 2006

National Academy of Sciences report on Surface Temperature Reconstructions, has served as an expert reviewer for the last three IPCC Assessment Reports (Working Groups I and II) and has testified before the U.S. Congress and committees of the Canadian House of Commons and Senate.

Roy W. Spencer, Ph.D., is a Principal Research Scientist at the University of Alabama in Huntsville. He holds a BS in Atmospheric and Oceanic Science from the University of Michigan, and MS and PhD in Meteorology from the University of Wisconsin -Madison. Spencer's published research has focused on monitoring of weather processes and climate change from Earth-orbiting satellites, as well as using simple physical models for diagnosing climate sensitivity and climate feedbacks. As a NASA Senior Scientist for Climate Studies he served as the U.S. Science Team Leader for the Advanced Microwave Scanning Radiometer (AMSR-E) flying on the Aqua satellite, providing our most accurate global measurements of sea surface temperatures, sea ice, and several other features of the climate system. Along with Dr. John Christy, Spencer developed the first satellite-based technique for global temperature monitoring, for which they received NASA's Exceptional Scientific Achievement Medal and the American Meteorological Society's Special Award.

								1973	1974	1975	
6	USH00015749	Tmax_adj	1973	2025	7	34.7442	-87.5997	164.6	32.4	32.4	31.5
10	USH00018024	Tmax_adj	1973	2025	7	33.4164	-86.135	136.6	31.9	32.3	31.2
3	USH00013160	Tmax_adj	1973	2025	7	32.8347	-88.1342	38.1	31.7	32.4	30.4
1	USH00011084	Tmax_adj	1973	2025	7	31.0581	-87.0547	25.9	33.5	34.2	32.5
2	USH00012813	Tmax_adj	1973	2025	7	30.5467	-87.8808	7	33	33.1	31.7
12	USH00018323	Tmax_adj	1973	2025	7	31.8075	-85.9722	165.2	33.2	32.4	30.6
13	USH00018380	Tmax_adj	1973	2025	7	33.2119	-87.6161	51.5	32.7	33.1	32.2
9	USH00017366	Tmax_adj	1973	2025	7	32.41	-87.0153	44.8	34.5	33.4	31.5
8	USH00017304	Tmax_adj	1973	2025	7	34.6736	-86.0536	187.5	31	30.8	30.4
11	USH00018178	Tmax_adj	1973	2025	7	31.5411	-87.8833	118.9	32.5	32.6	31.6
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5	USH00013816	Tmax_adj	1973	2025	7	31.87	-86.2542	132	32	32.2	30.2
15	USH00018469	Tmax_adj	1973	2025	7	34.5667	-85.6128	323.7	29.8	29.5	28.8
7	USH00017157	Tmax_adj	1973	2025	7	34.1736	-86.8133	243.8	31	31.1	30.5
14	USH00018438	Tmax_adj	1973	2025	7	32.0142	-85.7464	134.1	34	33.1	31.8
46	USH00032930	Tmax_adj	1973	2025	7	36.4261	-94.4481	384	30.8	34.6	31.5
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45	USH00032444	Tmax_adj	1973	2025	7	36.1006	-94.1744	387.1	30.4	33.3	30.3
48	USH00034756	Tmax_adj	1973	2025	7	34.5731	-94.2494	344.4	31	33.9	31.8
50	USH00035512	Tmax_adj	1973	2025	7	35.5125	-93.8683	253	32.3	34.6	32.4
55	USH00036928	Tmax_adj	1973	2025	7	35.3028	-93.6369	152.4	33.1	34.3	32.2
53	USH00035908	Tmax_adj	1973	2025	7	33.8203	-93.3878	93.9	32.4	33.4	32.9
51	USH00035754	Tmax_adj	1973	2025	7	34.2256	-92.0189	65.5	32.5	33.5	33
47	USH00034572	Tmax_adj	1973	2025	7	36.4947	-91.535	153	33.6	33.9	32.5
41	USH00030936	Tmax_adj	1973	2025	7	34.8822	-91.2153	56.4	33.5	33.6	31.6
44	USH00032356	Tmax_adj	1973	2025	7	36.4164	-93.7917	432.8	30.7	33.6	32.4
43	USH00031632	Tmax_adj	1973	2025	7	36.4197	-90.5858	91.4	34.1	34	32.6
49	USH00035186	Tmax_adj	1973	2025	7	35.6042	-91.2744	69.5	32.6	33.5	32
52	USH00035820	Tmax_adj	1973	2025	7	36.2639	-90.9681	96	33.4	33.9	32.6
54	USH00036253	Tmax_adj	1973	2025	7	33.81	-91.2703	45.7	32.5	32.4	32
18	USH00021248	Tmax_adj	1973	2025	7	36.1533	-109.5394	1710	30.4	29.5	29.9
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30	USH00027281	Tmax_adj	1973	2025	7	33.6731	-111.1508	672.1	36.8	36.2	36.3
28	USH00026353	Tmax_adj	1973	2025	7	31.9356	-109.8378	1326	31.1	29.9	31.8
38	USH00029287	Tmax_adj	1973	2025	7	33.9792	-112.7403	638.6	38.7	37.3	38.9
22	USH00023596	Tmax_adj	1973	2025	7	36.0528	-112.1503	2068	27.7	26.8	28
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36	USH00028815	Tmax_adj	1973	2025	7	32.2292	-110.9536	742.2	36.9	35.2	35
21	USH00023160	Tmax_adj	1973	2025	7	35.2681	-111.7428	2239	26.5	25.8	25
19	USH00021514	Tmax_adj	1973	2025	7	33.2058	-111.6819	434.3	39.1	37.1	38.6
25	USH00024849	Tmax_adj	1973	2025	7	36.8644	-111.6022	978.4	37.6	37.2	37.3
29	USH00026796	Tmax_adj	1973	2025	7	34.5706	-112.4322	1587	31.5	29.9	30.5
39	USH00029359	Tmax_adj	1973	2025	7	35.2406	-112.1903	2057	28	26.3	27.5
40	USH00029652	Tmax_adj	1973	2025	7	32.6114	-114.635	58.2	40.6	39.6	40.1
27	USH00026250	Tmax_adj	1973	2025	7	34.1547	-114.2897	128	42.3	40.7	41.5
35	USH00028619	Tmax_adj	1973	2025	7	31.7056	-110.0569	1405	32.3	31.4	31.1

16	USH00020080	Tmax_adj	1973	2025	7	32.3697	-112.86	539.5	40.6	38.9	39.8
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17	USH00021026	Tmax_adj	1973	2025	7	33.3761	-112.5828	271.3	41.6	40.4	41.3
20	USH00021614	Tmax_adj	1973	2025	7	34.3494	-111.6981	807.7	39	37.3	37.8
24	USH00024645	Tmax_adj	1973	2025	7	35.2	-114.0167	1079	35.5	33.7	34.6
32	USH00027390	Tmax_adj	1973	2025	7	32.815	-109.6808	900.4	36.5	35.2	35.3
31	USH00027370	Tmax_adj	1973	2025	7	33.08	-111.7417	391.7	41.7	39.8	40.7
37	USH00029271	Tmax_adj	1973	2025	7	33.8169	-109.9833	1561	32.4	32	31.8
26	USH00025512	Tmax_adj	1973	2025	7	33.4044	-110.87	1085	37.2	37.5	37
88	USH00046508	Tmax_adj	1973	2025	7	41.3089	-123.5317	122.8	33.1	30.9	30.6
109	USH00049866	Tmax_adj	1973	2025	7	41.7036	-122.6408	800.1	34.2	31.7	31.5
62	USH00041912	Tmax_adj	1973	2025	7	39.0911	-120.9481	725.4	31.2	29	29.4
59	USH00041614	Tmax_adj	1973	2025	7	41.5336	-120.1736	1423	32.5	30.9	30.5
72	USH00043761	Tmax_adj	1973	2025	7	41.8042	-123.3758	341.4	35.2	33.1	33.2
68	USH00042941	Tmax_adj	1973	2025	7	34.7042	-118.4275	932.7	31.5	31.1	31.3
92	USH00047195	Tmax_adj	1973	2025	7	39.9367	-120.9475	1042	32.2	30.2	29.8
100	USH00048758	Tmax_adj	1973	2025	7	39.1678	-120.1428	1899	24.7	23.4	23.7
106	USH00049490	Tmax_adj	1973	2025	7	40.7222	-122.9331	599.8	35.2	32.5	32.8
69	USH00043161	Tmax_adj	1973	2025	7	39.5092	-123.7567	37.5	15.6	17	16
74	USH00044232	Tmax_adj	1973	2025	7	36.7981	-118.2036	1204	38.5	37.3	37.5
86	USH00046399	Tmax_adj	1973	2025	7	34.4478	-119.2275	227.1	29	30.9	29.9
80	USH00045385	Tmax_adj	1973	2025	7	39.1458	-121.5853	17.4	35.7	34.4	34.1
58	USH00041048	Tmax_adj	1973	2025	7	32.9544	-115.5581	-30.5	40.8	40.2	41.3
63	USH00042239	Tmax_adj	1973	2025	7	32.9897	-116.5872	1414	27.6	26.3	27
89	USH00046719	Tmax_adj	1973	2025	7	34.1483	-118.1447	263.3	29.3	30.5	29.6
93	USH00047304	Tmax_adj	1973	2025	7	40.5175	-122.2986	151.5	38.3	35.2	35.5
104	USH00049200	Tmax_adj	1973	2025	7	38.3956	-121.9608	33.5	34.8	34	32.8
71	USH00043747	Tmax_adj	1973	2025	7	36.3219	-119.6356	74.7	35	35	35.5
75	USH00044259	Tmax_adj	1973	2025	7	33.7086	-116.2153	-6.4	40.9	39.2	39.8
85	USH00046175	Tmax_adj	1973	2025	7	33.6025	-117.8803	3	18.4	19.9	19
105	USH00049452	Tmax_adj	1973	2025	7	35.5975	-119.3531	105.2	36.3	36	35.5
96	USH00047902	Tmax_adj	1973	2025	7	34.4167	-119.6844	1.5	21.6	23.2	23
107	USH00049699	Tmax_adj	1973	2025	7	39.5231	-122.3058	71	34.3	33.2	31.8
90	USH00046730	Tmax_adj	1973	2025	7	35.6278	-120.6856	213.4	31.2	31.8	31.2
67	USH00042910	Tmax_adj	1973	2025	7	40.8097	-124.1603	6.1	14.2	15.4	14.9
82	USH00045983	Tmax_adj	1973	2025	7	41.3206	-122.3081	1094	29.5	25.8	26.2
70	USH00043257	Tmax_adj	1973	2025	7	36.78	-119.7194	101.5	35.9	35.9	35.3
81	USH00045532	Tmax_adj	1973	2025	7	37.2858	-120.5117	46.6	33.2	33	32.3
95	USH00047851	Tmax_adj	1973	2025	7	35.3056	-120.6639	96	23.4	24.2	22.4
101	USH00048839	Tmax_adj	1973	2025	7	35.0233	-118.7497	434.3	34.5	34.8	33.7
65	USH00042319	Tmax_adj	1973	2025	7	36.4622	-116.8669	-59.1	47.8	46.1	46.3
77	USH00044890	Tmax_adj	1973	2025	7	36.3817	-119.0264	156.4	37	36.4	36.2
66	USH00042728	Tmax_adj	1973	2025	7	38.3306	-120.6706	217.9	35.1	33.9	34.1
79	USH00045032	Tmax_adj	1973	2025	7	38.1061	-121.2878	12.2	31.5	31.7	30.7
94	USH00047306	Tmax_adj	1973	2025	7	34.0528	-117.1894	401.7	33.8	34.3	34.2
60	USH00041715	Tmax_adj	1973	2025	7	39.6911	-121.8211	56.4	35.6	33.1	32.6
57	USH00040924	Tmax_adj	1973	2025	7	33.6131	-114.5972	81.7	42.8	41.3	42.1

84	USH00046118	Tmax_adj	1973	2025	7	34.7675	-114.6189	271.3	43.3	41.8	42.6
108	USH00049855	Tmax_adj	1973	2025	7	37.75	-119.5897	1225	30.5	29.4	29.1
76	USH00044713	Tmax_adj	1973	2025	7	39.3183	-120.6392	1572	27.2	24.3	25.5
87	USH00046506	Tmax_adj	1973	2025	7	39.7458	-122.1997	77.4	34.4	32.9	32.4
78	USH00044997	Tmax_adj	1973	2025	7	37.6922	-121.7692	146.3	32.4	32	30.9
73	USH00043875	Tmax_adj	1973	2025	7	38.6175	-122.8731	32.9	30.6	29.7	28.9
61	USH00041758	Tmax_adj	1973	2025	7	32.64	-117.0858	17.1	21.5	23.1	21.4
98	USH00047965	Tmax_adj	1973	2025	7	38.4381	-122.6978	53	26.4	26.8	26.2
102	USH00049087	Tmax_adj	1973	2025	7	33.7025	-117.7539	71.6	28.7	30.9	28.9
64	USH00042294	Tmax_adj	1973	2025	7	38.535	-121.7761	18.3	34.3	33.8	32.2
103	USH00049122	Tmax_adj	1973	2025	7	39.1467	-123.2103	193.9	34.3	32.2	32.4
99	USH00048702	Tmax_adj	1973	2025	7	40.4167	-120.6631	1275	33	30.5	30.7
83	USH00046074	Tmax_adj	1973	2025	7	38.2778	-122.2647	10.7	26.5	26.9	26.5
56	USH00040693	Tmax_adj	1973	2025	7	37.8744	-122.2589	94.5	22.3	22.4	21.7
97	USH00047916	Tmax_adj	1973	2025	7	36.9906	-121.9911	39.6	22.3	22.2	22.6
91	USH00046826	Tmax_adj	1973	2025	7	38.2578	-122.6078	6.1	26.7	26.2	25.9
115	USH00052184	Tmax_adj	1973	2025	7	37.6742	-106.3247	2397	22.9	23.5	23
114	USH00051741	Tmax_adj	1973	2025	7	39.2425	-107.9631	1823	28.9	29.6	30
127	USH00055722	Tmax_adj	1973	2025	7	38.4858	-107.8792	1765	30.4	30.9	31.4
121	USH00053662	Tmax_adj	1973	2025	7	38.5258	-106.9675	2324	23.6	24.6	25.2
120	USH00053146	Tmax_adj	1973	2025	7	39.1653	-108.7331	1379	33	33.7	34
116	USH00052281	Tmax_adj	1973	2025	7	39.6261	-106.0353	2763	20.6	21.7	22.3
129	USH00057337	Tmax_adj	1973	2025	7	38.0858	-106.1444	2347	25	26.1	26
118	USH00053005	Tmax_adj	1973	2025	7	40.6147	-105.1314	1525	27.5	30.2	29.5
131	USH00058204	Tmax_adj	1973	2025	7	37.9492	-107.8733	2643	22.8	22.8	23.1
132	USH00058429	Tmax_adj	1973	2025	7	37.1786	-104.4869	1838	28.6	30.4	29
133	USH00059243	Tmax_adj	1973	2025	7	40.0583	-102.2189	1122	29.8	33.5	30.9
110	USH00050848	Tmax_adj	1973	2025	7	39.9919	-105.2667	1672	28.4	30.6	29.9
113	USH00051564	Tmax_adj	1973	2025	7	38.8236	-102.3486	1295	28.8	32	29.9
126	USH00055322	Tmax_adj	1973	2025	7	37.1742	-105.9392	2344	25.1	25.8	25.5
130	USH00057936	Tmax_adj	1973	2025	7	40.4883	-106.8233	2094	26.1	27.6	27.7
112	USH00051528	Tmax_adj	1973	2025	7	39.2203	-105.2783	2097	26.2	28.6	27.9
128	USH00057167	Tmax_adj	1973	2025	7	38.0392	-103.6933	1271	29.9	32	30.4
124	USH00054770	Tmax_adj	1973	2025	7	38.0936	-102.6306	1106	32.1	34.7	32.8
117	USH00052446	Tmax_adj	1973	2025	7	38.4775	-102.7808	1285	31.7	33.9	32.4
119	USH00053038	Tmax_adj	1973	2025	7	40.26	-103.8156	1329	30.9	34.3	33.1
122	USH00053951	Tmax_adj	1973	2025	7	37.7717	-107.1097	2758	24.3	24.4	25.3
111	USH00051294	Tmax_adj	1973	2025	7	38.46	-105.2256	1636	29.8	31.7	30.8
125	USH00054834	Tmax_adj	1973	2025	7	38.0636	-103.2153	1186	33.2	35.7	33.7
123	USH00054076	Tmax_adj	1973	2025	7	38.0494	-102.1236	1033	33.1	35.8	33.7
134	USH00062658	Tmax_adj	1973	2025	7	41.95	-73.3667	167.6	28	27.4	27.8
136	USH00067970	Tmax_adj	1973	2025	7	41.1247	-73.5475	57.9	27.2	28.7	27.5
137	USH00068138	Tmax_adj	1973	2025	7	41.795	-72.2283	198.1	26.6	26.2	26.2
135	USH00063207	Tmax_adj	1973	2025	7	41.3506	-72.0394	12.2	26.9	26.8	26.9
142	USH00079605	Tmax_adj	1973	2025	7	39.7739	-75.5414	82.3	28.2	28	27.2
141	USH00076410	Tmax_adj	1973	2025	7	39.6694	-75.7514	27.4	29.8	30.1	29.5
138	USH00072730	Tmax_adj	1973	2025	7	39.2583	-75.5167	9.1	28.5	28.8	27.8

140	USH00075915	Tmax_adj	1973	2025	7	38.8983	-75.425	10.7	30.6	30.7	29.1
139	USH00073595	Tmax_adj	1973	2025	7	38.8161	-75.5761	13.7	29.9	29.9	28.9
148	USH00082850	Tmax_adj	1973	2025	7	25.8489	-81.3897	1.5	32.9	31.4	31.8
162	USH00088758	Tmax_adj	1973	2025	7	30.3931	-84.3533	16.8	33.1	32.2	30.6
153	USH00083207	Tmax_adj	1973	2025	7	27.4622	-80.3539	7.6	31	29.9	30.4
146	USH00080611	Tmax_adj	1973	2025	7	26.6928	-80.6711	6.1	32.6	32	32.5
155	USH00084570	Tmax_adj	1973	2025	7	24.555	-81.7522	1.2	30.6	30.3	30.7
161	USH00087851	Tmax_adj	1973	2025	7	28.3378	-82.26	57.9	31.5	30.1	30.5
143	USH00080211	Tmax_adj	1973	2025	7	29.7258	-85.0206	6.1	31.7	30.6	29.5
151	USH00083163	Tmax_adj	1973	2025	7	26.1019	-80.2011	4.9	31.8	31.7	31.4
164	USH00088942	Tmax_adj	1973	2025	7	28.6242	-80.8158	1.5	31.3	29.9	31
147	USH00082220	Tmax_adj	1973	2025	7	30.7244	-86.0939	74.7	33.5	33.4	31.4
152	USH00083186	Tmax_adj	1973	2025	7	26.585	-81.8614	4.6	32.2	30.9	31.7
156	USH00084731	Tmax_adj	1973	2025	7	30.1853	-82.5942	59.4	32.3	31.1	30.4
163	USH00088824	Tmax_adj	1973	2025	7	28.1586	-82.7644	2.4	32.9	31.6	31.5
160	USH00087020	Tmax_adj	1973	2025	7	25.5819	-80.4361	3	31.5	31.4	31.4
149	USH00082915	Tmax_adj	1973	2025	7	29.755	-81.5389	1.5	31.1	30	29.6
145	USH00080478	Tmax_adj	1973	2025	7	27.8986	-81.8433	38.1	32.9	32	32.6
150	USH00082944	Tmax_adj	1973	2025	7	30.6589	-81.4636	4	32.9	31.4	31.2
157	USH00085275	Tmax_adj	1973	2025	7	30.4517	-83.4119	36.6	34	33.3	31.6
159	USH00086997	Tmax_adj	1973	2025	7	30.4781	-87.1869	34.1	32.8	32.5	31.2
144	USH00080228	Tmax_adj	1973	2025	7	27.2181	-81.8739	9.1	32.4	31.8	32.4
154	USH00084289	Tmax_adj	1973	2025	7	28.8031	-82.3125	12.2	34.1	33.1	32.9
158	USH00086414	Tmax_adj	1973	2025	7	29.0803	-82.0778	22.9	33	32	31.9
172	USH00093621	Tmax_adj	1973	2025	7	34.3006	-83.86	356.6	29.8	28.8	28.3
175	USH00095874	Tmax_adj	1973	2025	7	33.0831	-83.2497	112.2	32.5	32.2	30.3
180	USH00097847	Tmax_adj	1973	2025	7	32.13	-81.21	14	32.5	30.6	30.1
179	USH00097600	Tmax_adj	1973	2025	7	34.2453	-85.1514	200.9	30.7	31.7	31
182	USH00098703	Tmax_adj	1973	2025	7	31.4461	-83.4767	115.8	32.5	31.5	30
174	USH00094170	Tmax_adj	1973	2025	7	33.2842	-83.4681	74.7	32.6	32.3	30.3
167	USH00091340	Tmax_adj	1973	2025	7	31.1681	-81.5022	4	32.5	31.6	31.4
176	USH00095882	Tmax_adj	1973	2025	7	32.8703	-81.9672	59.4	32.8	31.7	30.7
186	USH00099186	Tmax_adj	1973	2025	7	31.2514	-82.3128	44.2	34.3	33.3	31.8
173	USH00093754	Tmax_adj	1973	2025	7	31.9881	-81.9522	61	34.2	33.1	31.8
166	USH00090586	Tmax_adj	1973	2025	7	30.8228	-84.6175	57.9	33.3	31.8	30.3
178	USH00097276	Tmax_adj	1973	2025	7	30.7836	-83.5692	56.4	32.9	32.2	31.1
183	USH00098740	Tmax_adj	1973	2025	7	34.5786	-83.3319	308.5	31.5	30.8	30.7
169	USH00092318	Tmax_adj	1973	2025	7	33.5972	-83.8436	234.4	31.1	31.1	30.2
171	USH00092966	Tmax_adj	1973	2025	7	32.2003	-83.2058	121.9	33.1	31.6	30.4
165	USH00090140	Tmax_adj	1973	2025	7	31.5339	-84.1489	54.9	33.7	32	29.3
177	USH00096335	Tmax_adj	1973	2025	7	33.4544	-84.8178	272.5	30.7	30.5	29.5
184	USH00099141	Tmax_adj	1973	2025	7	33.4028	-82.6222	149.4	33.2	32	30.7
170	USH00092475	Tmax_adj	1973	2025	7	34.5292	-83.99	475.5	30.9	30	30
187	USH00099291	Tmax_adj	1973	2025	7	32.8694	-85.1892	175.3	32.2	32	30.9
168	USH00091500	Tmax_adj	1973	2025	7	31.1903	-84.2036	53.3	31.1	31.4	29.5
181	USH00098535	Tmax_adj	1973	2025	7	32.6875	-84.5197	195.1	32.2	32	29.9
185	USH00099157	Tmax_adj	1973	2025	7	33.7264	-82.7058	189	32.2	31.2	30

309	USH00137979	Tmax_adj	1973	2025	7	42.6347	-95.1694	434.3	26.7	30.9	29.9
295	USH00132724	Tmax_adj	1973	2025	7	43.43	-94.8219	396.8	27	30.2	29.7
305	USH00135796	Tmax_adj	1973	2025	7	40.9486	-91.5647	222.5	28.5	30.6	29.9
291	USH00130600	Tmax_adj	1973	2025	7	41.8814	-92.2764	246.9	28.3	30.4	29.8
306	USH00135952	Tmax_adj	1973	2025	7	43.0453	-92.3122	349.9	26.4	28.5	27
300	USH00134063	Tmax_adj	1973	2025	7	41.3656	-93.6481	287.1	29.3	33.3	30.6
304	USH00135769	Tmax_adj	1973	2025	7	40.7053	-94.2428	359.7	28.5	32.8	30
290	USH00130133	Tmax_adj	1973	2025	7	43.0683	-94.3053	377.6	28.9	31.3	30.1
297	USH00132864	Tmax_adj	1973	2025	7	42.8503	-91.8158	344.4	27.3	29	28
310	USH00138296	Tmax_adj	1973	2025	7	42.0356	-92.5806	289.3	28.3	31.5	29.4
311	USH00138688	Tmax_adj	1973	2025	7	41.2828	-91.7067	210.3	29.7	31.8	31.2
307	USH00137147	Tmax_adj	1973	2025	7	43.43	-96.1686	411.5	28.5	32.8	31.1
298	USH00132977	Tmax_adj	1973	2025	7	43.2844	-93.6306	396.2	28.2	30.8	29.7
296	USH00132789	Tmax_adj	1973	2025	7	41.0211	-91.9553	225.6	28.9	31	30.3
302	USH00134735	Tmax_adj	1973	2025	7	42.7817	-96.1458	364.2	29.7	33.5	31.8
292	USH00131402	Tmax_adj	1973	2025	7	43.0775	-92.6714	309.1	28	30.4	29.3
294	USH00131635	Tmax_adj	1973	2025	7	41.7947	-90.2639	178.3	29.1	31.4	29.5
308	USH00137161	Tmax_adj	1973	2025	7	42.3969	-94.6292	364.2	28.3	32.2	30.2
301	USH00134142	Tmax_adj	1973	2025	7	42.5189	-93.2536	344.4	28.4	31.7	29.9
289	USH00130112	Tmax_adj	1973	2025	7	41.0656	-92.7867	268.2	29.2	32.3	30.9
303	USH00134894	Tmax_adj	1973	2025	7	41.6381	-95.7883	301.8	29	34	31.6
299	USH00132999	Tmax_adj	1973	2025	7	42.5836	-94.2006	347.5	28.1	32.2	29.5
293	USH00131533	Tmax_adj	1973	2025	7	40.7244	-95.0192	298.7	29.6	35.5	31.6
192	USH00101408	Tmax_adj	1973	2025	7	44.5733	-116.6753	807.7	31.8	30	31.9
195	USH00103143	Tmax_adj	1973	2025	7	46.0931	-115.5356	475.5	33.2	30.2	32.7
206	USH00105559	Tmax_adj	1973	2025	7	42.1492	-112.2872	1363	28.2	30	29.6
197	USH00103732	Tmax_adj	1973	2025	7	42.5872	-111.7275	1692	25.9	27.2	26.7
193	USH00101956	Tmax_adj	1973	2025	7	47.6789	-116.8017	650.1	30	26.2	29.6
213	USH00107264	Tmax_adj	1973	2025	7	48.9958	-116.5031	541	30.5	27.1	29.9
191	USH00100803	Tmax_adj	1973	2025	7	42.3353	-111.385	1818	27.5	28.7	28.6
214	USH00107386	Tmax_adj	1973	2025	7	48.3514	-116.8358	725.4	29.4	26.4	29.7
190	USH00100470	Tmax_adj	1973	2025	7	44.0425	-111.2739	1589	26	28.1	27.4
204	USH00105275	Tmax_adj	1973	2025	7	42.1231	-111.3139	1806	24.1	25.5	26
210	USH00106388	Tmax_adj	1973	2025	7	44.9647	-116.2864	1180	30.1	27.8	30.2
216	USH00108137	Tmax_adj	1973	2025	7	48.2928	-116.5578	640.1	28.8	26.2	28.3
188	USH00100010	Tmax_adj	1973	2025	7	42.9536	-112.8253	1343	28.9	30.2	30.5
189	USH00100448	Tmax_adj	1973	2025	7	43.5936	-115.9236	986	33.1	31.4	34.1
198	USH00104140	Tmax_adj	1973	2025	7	42.5972	-114.1378	1238	30.5	30.8	31.7
194	USH00102845	Tmax_adj	1973	2025	7	46.5022	-116.3217	303.3	33.5	31	32.5
209	USH00106305	Tmax_adj	1973	2025	7	43.6039	-116.5753	752.9	33.3	32.1	34.3
196	USH00103631	Tmax_adj	1973	2025	7	42.9403	-115.3231	751.6	35.5	34.8	36.6
200	USH00104670	Tmax_adj	1973	2025	7	42.7325	-114.5192	1140	31.2	30.9	32.8
212	USH00106891	Tmax_adj	1973	2025	7	44.0772	-116.9297	655.3	32.9	31.3	33.1
211	USH00106542	Tmax_adj	1973	2025	7	42.2342	-113.8981	1390	28.9	28.9	29.6
202	USH00104845	Tmax_adj	1973	2025	7	43.6842	-114.3603	1795	25.8	27.2	26.8
208	USH00106152	Tmax_adj	1973	2025	7	46.7244	-116.9608	810.8	30.5	27.4	28.8
207	USH00105685	Tmax_adj	1973	2025	7	44.5664	-113.8953	1539	30.7	31.7	31.4

215	USH00108080	Tmax_adj	1973	2025	7	45.1875	-113.9008	1198	31.7	31.2	30.4
199	USH00104295	Tmax_adj	1973	2025	7	42.3528	-114.5739	1379	29.1	29	30.2
203	USH00105241	Tmax_adj	1973	2025	7	46.3747	-117.0156	437.7	35.3	32.2	33.9
205	USH00105462	Tmax_adj	1973	2025	7	43.9181	-113.6317	1797	27.4	29.5	27.9
201	USH00104831	Tmax_adj	1973	2025	7	47.5311	-116.1267	707.1	31.6	29.1	32.2
221	USH00111436	Tmax_adj	1973	2025	7	39.4761	-88.1653	198.1	30.4	31.6	29
243	USH00116579	Tmax_adj	1973	2025	7	39.3722	-89.0242	213.4	30	31.3	28.8
238	USH00115901	Tmax_adj	1973	2025	7	42.0981	-89.9842	195.1	28.3	30.1	28.2
245	USH00116738	Tmax_adj	1973	2025	7	39.8058	-90.8236	198.1	28.3	30.8	28.8
224	USH00112348	Tmax_adj	1973	2025	7	41.8472	-89.5047	213.4	28.5	30.5	28.5
237	USH00115833	Tmax_adj	1973	2025	7	41.8039	-89.9744	183.8	28.4	30	27.8
228	USH00114108	Tmax_adj	1973	2025	7	39.15	-89.4833	192	29.4	30.9	28.8
230	USH00114442	Tmax_adj	1973	2025	7	39.7347	-90.1978	185.9	29.8	31.8	29.5
249	USH00118740	Tmax_adj	1973	2025	7	40.0839	-88.2403	219.8	29.9	31.2	28.8
234	USH00115515	Tmax_adj	1973	2025	7	38.0844	-88.5425	135.9	31.8	32.2	30.9
236	USH00115768	Tmax_adj	1973	2025	7	40.9247	-90.6392	227.1	27.6	29.3	27.8
219	USH00110338	Tmax_adj	1973	2025	7	41.7806	-88.3092	201.2	29.5	29.7	28.3
242	USH00116558	Tmax_adj	1973	2025	7	39.0028	-87.6225	140.2	29.7	30.5	29.2
220	USH00111280	Tmax_adj	1973	2025	7	39.2883	-89.8703	189.3	30.1	32.3	29.4
244	USH00116610	Tmax_adj	1973	2025	7	39.6356	-87.6933	207.3	30	31.2	29
233	USH00115326	Tmax_adj	1973	2025	7	42.2928	-88.6469	248.4	29.4	30.6	28.2
248	USH00118147	Tmax_adj	1973	2025	7	38.1167	-89.7167	163.1	31.1	32.4	30.6
226	USH00113335	Tmax_adj	1973	2025	7	41.1744	-90.0356	246.9	28.4	30	28.5
240	USH00116446	Tmax_adj	1973	2025	7	38.7003	-88.0819	146.3	31.2	32.2	30.5
252	USH00119354	Tmax_adj	1973	2025	7	39.4381	-88.5986	210.3	29.9	30.8	28.3
218	USH00110187	Tmax_adj	1973	2025	7	37.4814	-89.2344	195.1	30	31.2	28.9
225	USH00112483	Tmax_adj	1973	2025	7	37.9878	-89.1931	128	31.7	32.8	31.1
251	USH00119241	Tmax_adj	1973	2025	7	39.4411	-90.3789	176.8	29.5	32.2	29.5
246	USH00116910	Tmax_adj	1973	2025	7	40.8853	-88.6389	198.1	30	30.9	28.4
227	USH00113879	Tmax_adj	1973	2025	7	37.7408	-88.5244	111.3	32.8	33.1	32.4
229	USH00114198	Tmax_adj	1973	2025	7	40.4744	-87.6561	216.4	29.1	30.7	28.6
235	USH00115712	Tmax_adj	1973	2025	7	40.9125	-89.0339	228.6	29.8	32.2	29.2
217	USH00110072	Tmax_adj	1973	2025	7	41.1961	-90.7478	219.5	29.4	31.3	29.6
222	USH00112140	Tmax_adj	1973	2025	7	40.1389	-87.6483	170.1	29.6	30.9	28.6
231	USH00114823	Tmax_adj	1973	2025	7	40.5839	-90.9686	210.3	28.9	30.4	29.3
232	USH00115079	Tmax_adj	1973	2025	7	40.1514	-89.3381	177.7	29.2	30.9	28.5
223	USH00112193	Tmax_adj	1973	2025	7	39.8275	-88.9525	189	29.1	30	28.7
241	USH00116526	Tmax_adj	1973	2025	7	41.3283	-88.9106	160	28.6	30.6	28.2
250	USH00118916	Tmax_adj	1973	2025	7	41.5519	-89.5989	210.3	29.4	31	29
247	USH00117551	Tmax_adj	1973	2025	7	40.1158	-90.5608	201.2	29.1	31.4	30.3
239	USH00115943	Tmax_adj	1973	2025	7	38.3483	-88.8533	149.4	32.7	33.5	31.8
288	USH00129670	Tmax_adj	1973	2025	7	41.0267	-86.5867	210.3	27.6	28.7	27.4
281	USH00127935	Tmax_adj	1973	2025	7	38.9822	-85.9822	173.7	30	30	29.4
259	USH00121425	Tmax_adj	1973	2025	7	38.4833	-85.7	167.6	29.8	29.3	30.1
269	USH00125237	Tmax_adj	1973	2025	7	38.7369	-85.3942	140.2	30.8	30.8	30.8
283	USH00129080	Tmax_adj	1973	2025	7	38.7436	-85.0761	150.9	31	30.7	31
253	USH00120177	Tmax_adj	1973	2025	7	40.1122	-85.7175	257.6	28.2	29.2	28.6

254	USH00120200	Tmax_adj	1973	2025	7	41.6397	-84.9897	307.8	27.8	29.4	27.8
278	USH00127646	Tmax_adj	1973	2025	7	39.6042	-85.4528	292.6	29	29.8	29.1
276	USH00127482	Tmax_adj	1973	2025	7	41.0658	-86.2094	234.7	28.5	30.2	28.2
257	USH00121030	Tmax_adj	1973	2025	7	39.4239	-85.0128	192	30.5	31	31.4
273	USH00126705	Tmax_adj	1973	2025	7	38.5556	-86.4858	170.7	30.3	30.3	30.1
277	USH00127522	Tmax_adj	1973	2025	7	39.7594	-87.2269	211.2	29.2	30.4	28.6
263	USH00123418	Tmax_adj	1973	2025	7	41.5575	-85.8825	266.7	28.3	30.3	28.1
286	USH00129511	Tmax_adj	1973	2025	7	41.1947	-87.0578	202.7	29.2	31	30.1
282	USH00128036	Tmax_adj	1973	2025	7	38.5528	-86.7944	154.2	31.6	32	31.4
287	USH00129557	Tmax_adj	1973	2025	7	39.9961	-86.3542	289	28.1	28.8	27.5
258	USH00121229	Tmax_adj	1973	2025	7	39.8667	-85.1833	304.8	28.5	29.2	28.9
284	USH00129113	Tmax_adj	1973	2025	7	38.7386	-87.4878	137.2	31.6	32.8	31.4
274	USH00127125	Tmax_adj	1973	2025	7	38.3567	-87.5906	146.3	31.3	31.9	31
256	USH00120784	Tmax_adj	1973	2025	7	39.1742	-86.5214	253	29.8	30.8	29.9
270	USH00125337	Tmax_adj	1973	2025	7	40.58	-85.6586	240.8	29.5	30.8	29.5
280	USH00127875	Tmax_adj	1973	2025	7	38.6886	-85.785	173.7	31.6	31.4	31.6
255	USH00120676	Tmax_adj	1973	2025	7	40.6683	-84.9305	265.2	29	30.1	29.5
262	USH00122149	Tmax_adj	1973	2025	7	40.615	-86.6661	169.5	29.5	31.1	28.9
279	USH00127755	Tmax_adj	1973	2025	7	38.6111	-86.0889	243.8	30.6	30.5	30.3
265	USH00123527	Tmax_adj	1973	2025	7	39.7858	-85.7611	263.7	29.6	30.5	29.5
267	USH00124181	Tmax_adj	1973	2025	7	40.8556	-85.4981	221	29.1	30.5	29.2
275	USH00127298	Tmax_adj	1973	2025	7	40.9356	-87.1564	198.1	28.8	31	29.1
266	USH00124008	Tmax_adj	1973	2025	7	41.5422	-87.2881	195.1	28.5	29.1	27.8
271	USH00126001	Tmax_adj	1973	2025	7	37.9286	-87.8956	108.8	31.5	32.2	31.3
285	USH00129253	Tmax_adj	1973	2025	7	38.6489	-87.1989	152.4	31.7	31.8	31.1
268	USH00124837	Tmax_adj	1973	2025	7	41.6117	-86.7297	257.6	28.7	30.6	28.5
272	USH00126580	Tmax_adj	1973	2025	7	38.8894	-86.5519	198.1	30	30.3	29.8
260	USH00121747	Tmax_adj	1973	2025	7	39.1978	-85.9211	189.3	31	31.8	31.3
261	USH00121873	Tmax_adj	1973	2025	7	40.0028	-86.8011	256	29.3	30.3	29.5
264	USH00123513	Tmax_adj	1973	2025	7	39.6439	-86.8767	224	30	30.4	29.3
324	USH00144087	Tmax_adj	1973	2025	7	38.1964	-99.9181	740.7	33.3	35.5	33.1
325	USH00144464	Tmax_adj	1973	2025	7	37.9411	-101.2492	913.8	33.5	35.1	33.2
342	USH00148495	Tmax_adj	1973	2025	7	39.0297	-99.8831	749.8	32.1	35.3	32.4
335	USH00145906	Tmax_adj	1973	2025	7	39.8183	-100.5331	795.5	30.9	35	32.5
326	USH00144530	Tmax_adj	1973	2025	7	38.1817	-99.0994	608.1	31.7	34.5	31.7
338	USH00147093	Tmax_adj	1973	2025	7	39.7675	-101.8097	1025	31.5	35.7	33.7
313	USH00140365	Tmax_adj	1973	2025	7	37.1942	-99.7633	600.5	34.9	36.1	33.4
339	USH00147271	Tmax_adj	1973	2025	7	38.4819	-100.9189	905.3	32.6	34.5	32.2
329	USH00144695	Tmax_adj	1973	2025	7	37.0222	-100.9294	863.8	33	34.6	31.6
321	USH00143527	Tmax_adj	1973	2025	7	38.8586	-99.3358	612.6	33	35.7	33
314	USH00140405	Tmax_adj	1973	2025	7	39.5756	-95.1108	288	29.9	34.8	31.9
333	USH00145363	Tmax_adj	1973	2025	7	39.1247	-97.7047	402.9	33.3	37.6	33.7
341	USH00147542	Tmax_adj	1973	2025	7	39.7772	-98.7783	542.5	31.6	36.8	33
336	USH00145972	Tmax_adj	1973	2025	7	38.8875	-94.7603	321.6	30.7	34.7	32.3
334	USH00145856	Tmax_adj	1973	2025	7	39.7408	-99.8356	719.3	31.5	36.2	32.7
319	USH00142459	Tmax_adj	1973	2025	7	38.7278	-98.2267	466.3	34.1	37.5	33.8
316	USH00141740	Tmax_adj	1973	2025	7	37.1764	-94.8397	275.8	32	35.3	33.6

328	USH00144588	Tmax_adj	1973	2025	7	39.3256	-94.9189	265.2	31.4	35.8	33
330	USH00144972	Tmax_adj	1973	2025	7	39.1972	-96.5814	324.6	31.5	35.2	32.8
340	USH00147305	Tmax_adj	1973	2025	7	37.1317	-96.1861	274.3	33.1	36.9	34
323	USH00143954	Tmax_adj	1973	2025	7	37.2364	-95.7003	245.4	32.5	35.9	33
315	USH00141704	Tmax_adj	1973	2025	7	37.2706	-99.3272	634.9	32.1	35.5	31.8
331	USH00145152	Tmax_adj	1973	2025	7	38.3772	-97.6097	463.3	33.3	37.6	33.9
327	USH00144559	Tmax_adj	1973	2025	7	38.9583	-95.2514	306.3	31.8	35.8	32.9
332	USH00145173	Tmax_adj	1973	2025	7	37.2767	-98.58	448.1	34.1	36.7	33.4
318	USH00142401	Tmax_adj	1973	2025	7	37.8183	-96.8444	393.2	33.6	36.7	33
320	USH00142835	Tmax_adj	1973	2025	7	37.8428	-94.7086	257.6	32.9	35.9	32.7
317	USH00141867	Tmax_adj	1973	2025	7	38.6758	-96.5097	402.3	31.8	36.6	32.6
312	USH00140264	Tmax_adj	1973	2025	7	37.155	-98.0283	408.4	34.4	37.4	33.3
322	USH00143810	Tmax_adj	1973	2025	7	39.6703	-95.5225	313.9	31	35.4	32.7
337	USH00146128	Tmax_adj	1973	2025	7	38.6133	-95.2808	280.1	32.6	36.1	33.3
343	USH00150254	Tmax_adj	1973	2025	7	38.4536	-82.6131	170.7	29.8	29.4	29
355	USH00158714	Tmax_adj	1973	2025	7	38.6586	-84.6106	286.5	28.8	28.9	29.2
348	USH00153028	Tmax_adj	1973	2025	7	38.2022	-84.8825	140.8	30	29.5	30.2
353	USH00157324	Tmax_adj	1973	2025	7	38.2	-85.2	222.5	30.2	30.2	30.8
354	USH00158709	Tmax_adj	1973	2025	7	36.7333	-84.15	286.5	27.3	28.4	27.5
350	USH00153762	Tmax_adj	1973	2025	7	37.7558	-87.6456	136.9	30.5	30.8	29.9
344	USH00150381	Tmax_adj	1973	2025	7	36.8825	-83.8819	301.8	30.9	31.1	31.9
347	USH00152791	Tmax_adj	1973	2025	7	38.1167	-83.55	207.3	29.1	28.5	29.2
349	USH00153430	Tmax_adj	1973	2025	7	37.2592	-85.5022	179.8	31	31.3	31.9
345	USH00150619	Tmax_adj	1973	2025	7	37.5733	-84.2908	326.1	29.1	29.5	30
346	USH00150909	Tmax_adj	1973	2025	7	36.9647	-86.4239	160.9	31.8	32.2	32
351	USH00153994	Tmax_adj	1973	2025	7	36.8492	-87.5206	158.5	31.2	31.2	31.5
352	USH00154703	Tmax_adj	1973	2025	7	37.5108	-86.2892	189	31	30.3	30.6
369	USH00166664	Tmax_adj	1973	2025	7	29.9167	-90.1303	6.1	32.6	31.2	30.9
359	USH00160549	Tmax_adj	1973	2025	7	30.5372	-91.1469	19.5	33.6	32.8	32.3
372	USH00169013	Tmax_adj	1973	2025	7	29.7758	-90.78	4.6	33.1	31.8	31
366	USH00164407	Tmax_adj	1973	2025	7	29.5858	-90.7303	4.6	33	32.3	31.5
368	USH00165026	Tmax_adj	1973	2025	7	30.205	-91.9875	11.6	33	32.5	31.7
362	USH00162151	Tmax_adj	1973	2025	7	30.5272	-90.1114	12.2	32.7	32	31.2
363	USH00162534	Tmax_adj	1973	2025	7	30.0717	-91.0275	9.1	32.5	30.6	31.2
365	USH00163800	Tmax_adj	1973	2025	7	30.4192	-92.0439	16.8	32.8	32	31.2
361	USH00161411	Tmax_adj	1973	2025	7	32.5133	-92.3478	54.9	33.8	33.4	32.7
373	USH00169806	Tmax_adj	1973	2025	7	32.0994	-91.7019	24.4	34.3	33.6	32.8
370	USH00167344	Tmax_adj	1973	2025	7	32.9078	-93.7981	88.4	33	33.3	32.9
364	USH00163313	Tmax_adj	1973	2025	7	29.8233	-91.5442	3.7	32	31.8	30.9
367	USH00164700	Tmax_adj	1973	2025	7	30.2003	-92.6642	7.6	32.7	31.8	30.7
356	USH00160098	Tmax_adj	1973	2025	7	31.3206	-92.4611	26.5	33.6	33.2	32.2
358	USH00160537	Tmax_adj	1973	2025	7	32.7689	-92.0056	45.7	32.6	32.5	32
371	USH00168163	Tmax_adj	1973	2025	7	31.9497	-91.2336	23.8	34.5	34.3	33.5
357	USH00160205	Tmax_adj	1973	2025	7	30.7094	-90.525	51.8	34.6	33.7	32.8
360	USH00161287	Tmax_adj	1973	2025	7	30.9594	-92.1786	24.4	33.6	33.1	32.4
407	USH00195246	Tmax_adj	1973	2025	7	41.6333	-70.9333	21.3	26.7	26.9	27.3
403	USH00190535	Tmax_adj	1973	2025	7	42.4833	-71.2833	48.8	27.7	27.6	28.6

402	USH00190120	Tmax_adj	1973	2025	7	42.3861	-72.5375	45.7	28.1	27.5	28.1
405	USH00193213	Tmax_adj	1973	2025	7	42.145	-73.4172	249	28	27.1	28
406	USH00194105	Tmax_adj	1973	2025	7	42.6992	-71.1658	15.2	28.5	28.4	29.4
409	USH00196681	Tmax_adj	1973	2025	7	42.05	-70.1833	6.1	25.6	26.1	26.7
412	USH00198757	Tmax_adj	1973	2025	7	42.1608	-71.2458	50.3	28.6	28.3	29
411	USH00198367	Tmax_adj	1973	2025	7	41.9003	-71.0658	6.1	27.8	28.1	28.7
404	USH00190736	Tmax_adj	1973	2025	7	42.2122	-71.1136	192	26.9	27.1	28
410	USH00196783	Tmax_adj	1973	2025	7	42.5242	-71.1264	27.4	28.2	28.1	28.5
408	USH00196486	Tmax_adj	1973	2025	7	41.9819	-70.6961	13.7	28.1	28.7	28.8
413	USH00199316	Tmax_adj	1973	2025	7	42.1333	-71.4333	64	29.3	29.7	29.3
388	USH00181750	Tmax_adj	1973	2025	7	39.2167	-76.0519	12.2	28.6	28.7	27.9
394	USH00185985	Tmax_adj	1973	2025	7	39.2739	-75.87	9.1	29.7	29.4	28.6
401	USH00189750	Tmax_adj	1973	2025	7	39.3333	-76.8667	140.2	30.4	30.1	29.3
400	USH00189440	Tmax_adj	1973	2025	7	39.55	-76.9667	233.2	29	29.2	28.4
398	USH00187806	Tmax_adj	1973	2025	7	38.7144	-76.1897	3	30	29.9	28.8
399	USH00188000	Tmax_adj	1973	2025	7	38.365	-75.5892	3	29.3	30.1	28.7
392	USH00185111	Tmax_adj	1973	2025	7	39.0847	-76.9003	121.9	29.3	29.9	29.4
391	USH00183675	Tmax_adj	1973	2025	7	38.9689	-76.8042	45.7	30.5	30.3	29.5
389	USH00182282	Tmax_adj	1973	2025	7	39.6419	-78.7561	222.5	30.8	30	29.3
393	USH00185718	Tmax_adj	1973	2025	7	39.2811	-76.61	6.1	32	32.4	31.3
386	USH00180700	Tmax_adj	1973	2025	7	39.0303	-76.9314	44.2	29.7	29.9	28.9
387	USH00181385	Tmax_adj	1973	2025	7	38.5667	-76.0667	3	30.5	30.1	29.1
390	USH00182523	Tmax_adj	1973	2025	7	38.8833	-75.8	14.9	30.7	30.6	29.1
396	USH00186770	Tmax_adj	1973	2025	7	38.6833	-76.6667	48.8	30.1	30	29.1
397	USH00187330	Tmax_adj	1973	2025	7	38.2122	-75.6822	6.1	30.6	31.4	30.5
395	USH00186620	Tmax_adj	1973	2025	7	39.4133	-79.4	737.6	27.5	27.2	27
382	USH00175304	Tmax_adj	1973	2025	7	45.6503	-68.705	109.7	26.9	24.6	26.8
385	USH00179891	Tmax_adj	1973	2025	7	45.1572	-67.4044	42.7	27.5	25.6	26.5
380	USH00173944	Tmax_adj	1973	2025	7	46.2061	-67.8417	118.9	26.8	24.4	27.2
381	USH00174566	Tmax_adj	1973	2025	7	44.1	-70.2167	54.9	26.9	25.4	27.2
375	USH00170814	Tmax_adj	1973	2025	7	45.6603	-69.812	323.1	25.7	23.1	26.7
384	USH00176937	Tmax_adj	1973	2025	7	46.6539	-68.0089	182.6	27	24.7	27.4
383	USH00176905	Tmax_adj	1973	2025	7	43.6497	-70.3003	13.7	26.3	25.1	26.4
377	USH00172426	Tmax_adj	1973	2025	7	44.9067	-66.9919	25.9	24.2	22.5	22.9
378	USH00172765	Tmax_adj	1973	2025	7	44.6889	-70.1567	128	27	25.4	27.4
379	USH00173046	Tmax_adj	1973	2025	7	44.2203	-69.7889	42.7	27.3	25.8	27.5
374	USH00170100	Tmax_adj	1973	2025	7	44.3739	-68.2592	143.3	25.9	24.5	25.8
376	USH00171628	Tmax_adj	1973	2025	7	44.9197	-69.2417	90.5	27.8	26.8	27.2
432	USH00205662	Tmax_adj	1973	2025	7	43.5858	-84.7697	242.6	27.1	27.7	27.2
434	USH00205816	Tmax_adj	1973	2025	7	46.3133	-85.5106	259.1	22.8	23.3	24.4
425	USH00203632	Tmax_adj	1973	2025	7	43.6747	-86.4239	234.7	26.4	26.8	26.9
420	USH00201486	Tmax_adj	1973	2025	7	46.3414	-86.9242	265.2	24.9	25.2	26
426	USH00203823	Tmax_adj	1973	2025	7	41.9353	-84.6411	329.2	27.7	28.9	28
431	USH00205650	Tmax_adj	1973	2025	7	42.6083	-82.8183	176.8	26.6	29	26.9
427	USH00204090	Tmax_adj	1973	2025	7	45.7858	-88.0842	326.4	26	26.6	27.3
428	USH00204104	Tmax_adj	1973	2025	7	46.4656	-90.1892	435.9	24	25.5	25.4
418	USH00200779	Tmax_adj	1973	2025	7	43.7072	-85.4822	283.5	27	27.7	27.4

424	USH00202737	Tmax_adj	1973	2025	7	45.6667	-86.7167	227.1	23.6	22.7	23.7
421	USH00201492	Tmax_adj	1973	2025	7	45.6528	-84.4725	179.2	25.9	25.7	25.5
417	USH00200230	Tmax_adj	1973	2025	7	42.2947	-83.7108	274.3	27.8	28.7	28.3
435	USH00206300	Tmax_adj	1973	2025	7	43.0161	-84.18	222.5	28.1	28.7	28.1
436	USH00207690	Tmax_adj	1973	2025	7	42.4011	-86.2833	189	24.9	25.6	24.6
414	USH00200032	Tmax_adj	1973	2025	7	41.9164	-84.0158	231.6	28	29.8	28.5
423	USH00202423	Tmax_adj	1973	2025	7	44.2839	-83.5036	178.6	26.7	25.7	26.9
437	USH00207812	Tmax_adj	1973	2025	7	46.0556	-88.6275	442	25.1	26	26.2
419	USH00201439	Tmax_adj	1973	2025	7	46.5192	-87.9858	487.4	22.7	24.5	25.1
429	USH00204244	Tmax_adj	1973	2025	7	42.2833	-85.6	289.6	28.2	29.7	28.7
415	USH00200128	Tmax_adj	1973	2025	7	42.58	-85.7892	228.6	26.8	29.2	27.5
422	USH00201675	Tmax_adj	1973	2025	7	41.9622	-84.9925	299.9	27.5	28.9	27.4
416	USH00200146	Tmax_adj	1973	2025	7	43.3864	-84.6492	224	28.4	28.8	29.1
430	USH00205434	Tmax_adj	1973	2025	7	43.6089	-84.2011	195.1	26.4	26.2	26.3
433	USH00205690	Tmax_adj	1973	2025	7	46.4122	-86.6625	207.3	23.5	24.3	25.2
467	USH00218419	Tmax_adj	1973	2025	7	47.0258	-91.6653	190.5	21.9	22.7	23.3
443	USH00211630	Tmax_adj	1973	2025	7	46.7047	-92.5253	385.6	24.2	26.2	26.2
450	USH00213303	Tmax_adj	1973	2025	7	47.2436	-93.4975	399.3	25.4	28	27.8
464	USH00217087	Tmax_adj	1973	2025	7	48.8486	-95.7675	319.1	25	27.3	26.3
452	USH00214652	Tmax_adj	1973	2025	7	47.2467	-94.2228	396.8	24.7	27.8	27.1
445	USH00212645	Tmax_adj	1973	2025	7	47.4581	-92.5303	440.4	24	26.8	26.3
466	USH00217460	Tmax_adj	1973	2025	7	46.7953	-93.3211	376.1	24.4	27.4	26.9
451	USH00214106	Tmax_adj	1973	2025	7	47.2256	-95.1919	454.2	25.1	28.2	27.3
442	USH00211465	Tmax_adj	1973	2025	7	44.8	-93.5833	219.5	28.3	30.5	30.1
455	USH00215435	Tmax_adj	1973	2025	7	44.8831	-93.2289	265.8	28.5	30.6	30.2
462	USH00216547	Tmax_adj	1973	2025	7	46.6694	-94.1089	381	26	28.7	28.3
441	USH00210515	Tmax_adj	1973	2025	7	48.7094	-94.5869	323.7	27	28.6	28.9
468	USH00218618	Tmax_adj	1973	2025	7	47.0744	-94.57	429.8	24.6	27.6	27.3
453	USH00215175	Tmax_adj	1973	2025	7	47.6308	-93.6522	422.8	25	27.1	27
461	USH00216360	Tmax_adj	1973	2025	7	46.9006	-95.0678	437.1	25.7	28.6	28.5
448	USH00212916	Tmax_adj	1973	2025	7	47.5636	-95.7244	399.3	25.1	28.2	27.4
458	USH00215638	Tmax_adj	1973	2025	7	45.5903	-95.8747	347.5	27	29.5	29
454	USH00215400	Tmax_adj	1973	2025	7	45.1219	-95.9269	310.9	28.9	31.6	32
440	USH00210252	Tmax_adj	1973	2025	7	48.3311	-96.8253	258.2	27.5	31.5	29.8
446	USH00212698	Tmax_adj	1973	2025	7	43.6447	-94.4656	361.8	27.9	30.9	30.7
447	USH00212737	Tmax_adj	1973	2025	7	44.6697	-93.17	298.7	28	30.8	29.9
457	USH00215615	Tmax_adj	1973	2025	7	45.8775	-93.3147	310.3	27.7	30.2	29.6
438	USH00210018	Tmax_adj	1973	2025	7	47.2992	-96.5161	276.5	26.7	30.3	29.7
449	USH00213290	Tmax_adj	1973	2025	7	43.7047	-92.5644	411.5	27.3	29.4	28.8
444	USH00212142	Tmax_adj	1973	2025	7	46.8372	-95.8372	413	27.8	29.5	27.7
469	USH00219046	Tmax_adj	1973	2025	7	43.7692	-94.1872	338.3	28.5	31.4	31.4
470	USH00219249	Tmax_adj	1973	2025	7	44.2992	-92.6661	300.2	28.2	30.8	29.4
439	USH00210075	Tmax_adj	1973	2025	7	43.6064	-93.3019	374.9	28.5	31.2	29.9
465	USH00217405	Tmax_adj	1973	2025	7	44.3222	-93.9656	259.1	28.6	31.7	31.2
459	USH00215887	Tmax_adj	1973	2025	7	44.3006	-94.4897	271.3	27.4	30.5	30
460	USH00216152	Tmax_adj	1973	2025	7	44.7628	-94.9297	335.3	28	31.7	31
456	USH00215563	Tmax_adj	1973	2025	7	44.9364	-95.7536	300.2	28.5	31.5	32

463	USH00216565	Tmax_adj	1973	2025	7	44.0139	-96.3258	519.7	29	32.8	32.3
505	USH00231037	Tmax_adj	1973	2025	7	39.4247	-93.1331	201.8	29.7	33.3	31.8
510	USH00232809	Tmax_adj	1973	2025	7	37.7922	-90.4103	282.9	31.2	33.2	30.8
528	USH00238725	Tmax_adj	1973	2025	7	38.835	-91.1386	252.1	30.2	33.1	30.4
504	USH00230856	Tmax_adj	1973	2025	7	39.3447	-91.1711	270.4	29.5	29.8	29.5
511	USH00234271	Tmax_adj	1973	2025	7	38.585	-92.1825	204.2	31.4	33.8	32.2
512	USH00234705	Tmax_adj	1973	2025	7	37.4994	-94.2697	298.7	32	35	32.6
513	USH00234825	Tmax_adj	1973	2025	7	37.685	-92.6936	389.8	32	33.8	32.2
520	USH00235834	Tmax_adj	1973	2025	7	37.1528	-92.2636	442	31.1	33.3	30.9
517	USH00235253	Tmax_adj	1973	2025	7	37.3036	-89.9664	118.9	32.7	33.8	31.3
522	USH00237263	Tmax_adj	1973	2025	7	37.9572	-91.7758	355.7	32.7	34.2	33.2
506	USH00231364	Tmax_adj	1973	2025	7	36.1686	-89.6642	82.3	33	34.1	32
514	USH00234850	Tmax_adj	1973	2025	7	38.8803	-94.3358	304.8	30	33.3	31.3
518	USH00235541	Tmax_adj	1973	2025	7	39.1756	-91.8861	244.4	31.2	34	32.1
524	USH00238051	Tmax_adj	1973	2025	7	39.9714	-91.8872	210.3	29.7	32.4	31.1
523	USH00237963	Tmax_adj	1973	2025	7	40.2472	-93.7158	266.7	29.2	33.3	31.8
516	USH00235027	Tmax_adj	1973	2025	7	37.3908	-93.9489	326.1	32.1	34.9	32.8
521	USH00235976	Tmax_adj	1973	2025	7	36.8639	-94.36	308.2	32	35.7	32.6
507	USH00231711	Tmax_adj	1973	2025	7	38.395	-93.7711	234.7	32.1	34.7	33.7
509	USH00232289	Tmax_adj	1973	2025	7	36.6206	-90.8125	88.1	33.7	34.7	33.3
526	USH00238466	Tmax_adj	1973	2025	7	38.2581	-93.3989	192.6	32.2	34.1	33.3
508	USH00231822	Tmax_adj	1973	2025	7	40.2394	-94.6833	337.7	29.4	34.8	31.6
525	USH00238223	Tmax_adj	1973	2025	7	38.9664	-93.4194	205.7	32.1	34.7	33.1
527	USH00238523	Tmax_adj	1973	2025	7	40.475	-93.0031	323.1	30	33	31.4
503	USH00230204	Tmax_adj	1973	2025	7	38.1869	-94.0281	259.7	32.1	34.6	33.5
515	USH00234904	Tmax_adj	1973	2025	7	39.1828	-93.855	251.5	31.6	35.3	33.9
519	USH00235671	Tmax_adj	1973	2025	7	39.4194	-92.4369	262.1	30.6	33.8	32.6
494	USH00227128	Tmax_adj	1973	2025	7	30.8408	-89.5453	95.4	32.2	31.8	30
488	USH00225247	Tmax_adj	1973	2025	7	33.1353	-89.0711	177.1	32.8	33	31.7
487	USH00224939	Tmax_adj	1973	2025	7	31.6756	-89.1236	68.6	33.8	33.4	32.6
480	USH00222094	Tmax_adj	1973	2025	7	31.9461	-90.3775	148.4	33.8	33.8	32.8
496	USH00228374	Tmax_adj	1973	2025	7	33.4692	-88.7822	56.4	33.2	32.8	32.2
486	USH00224776	Tmax_adj	1973	2025	7	33.0583	-89.5797	125	31.9	31.4	31.2
478	USH00221880	Tmax_adj	1973	2025	7	33.4678	-88.3847	44.2	33.6	33.6	33.5
479	USH00221962	Tmax_adj	1973	2025	7	34.8792	-88.6178	117.3	31.3	31.7	30.3
485	USH00224173	Tmax_adj	1973	2025	7	34.8219	-89.4347	147.2	32.8	32.1	31.9
492	USH00226718	Tmax_adj	1973	2025	7	30.3964	-88.4781	3.7	32.6	31.6	30.8
473	USH00220955	Tmax_adj	1973	2025	7	34.6628	-88.5797	149.4	31.4	31	30.3
481	USH00223107	Tmax_adj	1973	2025	7	32.36	-89.4208	137.2	32.1	32	31.3
489	USH00225987	Tmax_adj	1973	2025	7	31.5519	-90.1058	58.2	36.2	35.1	35.3
477	USH00221865	Tmax_adj	1973	2025	7	31.2503	-89.8361	45.7	34.2	33.5	32.8
495	USH00227132	Tmax_adj	1973	2025	7	31.985	-90.9719	36.6	34	33.9	32.7
501	USH00229793	Tmax_adj	1973	2025	7	31.0928	-91.2328	121.9	34.3	33.6	32.8
475	USH00221389	Tmax_adj	1973	2025	7	32.6714	-90.0361	76.2	33.6	33	32
491	USH00226177	Tmax_adj	1973	2025	7	31.5889	-91.3408	59.4	32.9	33.2	32
474	USH00221094	Tmax_adj	1973	2025	7	31.5447	-90.4581	132.6	33	33	31.9
482	USH00223605	Tmax_adj	1973	2025	7	33.3592	-91.06	38.1	32	32.4	32.2

483	USH00223887	Tmax_adj	1973	2025	7	31.2547	-89.3392	117.3	33.7	33.4	32.4
498	USH00229400	Tmax_adj	1973	2025	7	34.1583	-89.6311	94.5	32.5	32.3	32.6
502	USH00229860	Tmax_adj	1973	2025	7	32.9028	-90.3817	32.6	35.3	34.8	34.2
471	USH00220021	Tmax_adj	1973	2025	7	33.83	-88.5214	60.4	32.7	32.8	32.2
497	USH00229079	Tmax_adj	1973	2025	7	34.3803	-89.5358	124.4	33.5	33.5	32.4
499	USH00229426	Tmax_adj	1973	2025	7	30.2947	-89.3831	2.4	33.3	31.7	30.8
500	USH00229439	Tmax_adj	1973	2025	7	31.6772	-88.6708	61	34.3	33.8	33.1
493	USH00227111	Tmax_adj	1973	2025	7	34.1381	-88.9983	123.4	32.3	31.6	31.6
472	USH00220488	Tmax_adj	1973	2025	7	34.3061	-89.9806	67.1	33.3	33.6	33.1
490	USH00226009	Tmax_adj	1973	2025	7	33.4519	-90.5094	35.7	32.9	32.5	32.9
476	USH00221707	Tmax_adj	1973	2025	7	34.1864	-90.5572	52.7	34.4	35	34
484	USH00223975	Tmax_adj	1973	2025	7	34.8161	-89.9853	115.8	33.4	33.4	32
547	USH00243885	Tmax_adj	1973	2025	7	46.2553	-114.1642	1081	28.2	26.5	28.1
537	USH00242409	Tmax_adj	1973	2025	7	45.2128	-112.6447	1594	26.6	27.5	26
554	USH00245015	Tmax_adj	1973	2025	7	48.4036	-115.5392	638.9	33.1	29.2	32
550	USH00244345	Tmax_adj	1973	2025	7	45.9228	-108.2444	924.8	28.8	31.5	30.7
565	USH00247286	Tmax_adj	1973	2025	7	47.315	-114.0983	883.9	30.5	27.9	29.8
536	USH00242173	Tmax_adj	1973	2025	7	48.6033	-112.3753	1170	27.3	27.3	25.6
562	USH00246472	Tmax_adj	1973	2025	7	46.3158	-113.3	1606	27.2	27	27.1
529	USH00240199	Tmax_adj	1973	2025	7	46.1314	-112.9569	1609	26.6	27	25.7
543	USH00243139	Tmax_adj	1973	2025	7	48.7783	-114.8997	914.4	28.8	26.3	29.6
560	USH00245761	Tmax_adj	1973	2025	7	47.0575	-109.9514	1311	27.4	28.7	27
571	USH00248857	Tmax_adj	1973	2025	7	44.65	-111.1	2030	25.5	27.3	26.6
539	USH00242793	Tmax_adj	1973	2025	7	45.3394	-111.7111	1510	27.8	29	27.6
570	USH00248597	Tmax_adj	1973	2025	7	45.2925	-111.9481	1760	25.5	27.1	26.2
566	USH00247318	Tmax_adj	1973	2025	7	47.3033	-115.0908	810.8	31	29.5	31.3
544	USH00243558	Tmax_adj	1973	2025	7	48.2139	-106.6214	696.5	28.4	30.1	29.6
533	USH00241552	Tmax_adj	1973	2025	7	47.2194	-111.71	1024	30.3	30.7	29.2
553	USH00244558	Tmax_adj	1973	2025	7	48.3042	-114.2636	901.3	30.4	27.9	30.4
540	USH00243013	Tmax_adj	1973	2025	7	46.8511	-108.3133	954.9	28.4	30.2	28.3
549	USH00244055	Tmax_adj	1973	2025	7	46.6056	-111.9636	1167	31.3	31.6	29.6
532	USH00241044	Tmax_adj	1973	2025	7	45.6622	-111.0453	1498	26.6	28.7	26.6
564	USH00246918	Tmax_adj	1973	2025	7	45.1797	-109.2497	1718	25.7	27	25.9
551	USH00244364	Tmax_adj	1973	2025	7	45.9353	-107.1375	944.9	28.5	30.6	29.5
572	USH00248930	Tmax_adj	1973	2025	7	46.5436	-110.9	1536	26.8	27.9	26.5
561	USH00246157	Tmax_adj	1973	2025	7	45.4856	-111.6325	1446	28.7	30	28.4
530	USH00240364	Tmax_adj	1973	2025	7	47.4931	-112.3964	1241	26.8	26.2	25
548	USH00244038	Tmax_adj	1973	2025	7	44.8667	-111.3392	1978	24.1	26.3	25.2
534	USH00241722	Tmax_adj	1973	2025	7	48.5883	-109.2256	737.6	29.2	29.7	29.7
546	USH00243751	Tmax_adj	1973	2025	7	47.4733	-111.3822	1117	30.9	31.4	29.9
531	USH00240780	Tmax_adj	1973	2025	7	45.8328	-109.9503	1250	28.9	31.1	29.7
568	USH00248501	Tmax_adj	1973	2025	7	48.3089	-112.2511	1161	28	27.5	27.4
556	USH00245338	Tmax_adj	1973	2025	7	48.3939	-107.7286	680	29.7	30	31.3
555	USH00245080	Tmax_adj	1973	2025	7	45.4836	-110.5689	1484	27.6	28.2	27.6
559	USH00245690	Tmax_adj	1973	2025	7	46.4267	-105.8825	799.8	30	33.3	31.9
552	USH00244522	Tmax_adj	1973	2025	7	47.3144	-106.9103	798.6	30.3	31.3	30.4
541	USH00243089	Tmax_adj	1973	2025	7	48.7778	-107.4536	792.2	29.7	29.3	30.4

535	USH00241737	Tmax_adj	1973	2025	7	47.8206	-112.1919	1172	28.9	29.1	27.8
538	USH00242689	Tmax_adj	1973	2025	7	45.8903	-104.5461	1044	27.7	30.6	28.9
557	USH00245572	Tmax_adj	1973	2025	7	48.4828	-104.4514	591.9	28.9	30.6	29.8
542	USH00243110	Tmax_adj	1973	2025	7	48.4983	-109.7972	796.4	29.5	30	29.1
558	USH00245668	Tmax_adj	1973	2025	7	46.7614	-104.9619	757.1	30.1	32.9	31.4
545	USH00243581	Tmax_adj	1973	2025	7	47.1064	-104.7183	632.8	29	31.5	30.1
563	USH00246601	Tmax_adj	1973	2025	7	46.4178	-104.5164	847.3	29.3	31.3	29.8
567	USH00247382	Tmax_adj	1973	2025	7	47.4536	-104.3378	602	29.8	33	31
569	USH00248569	Tmax_adj	1973	2025	7	47.88	-105.3686	696.2	29.9	32	31.2
761	USH00318694	Tmax_adj	1973	2025	7	36.3919	-81.3039	876.3	25.7	24.9	25
738	USH00312635	Tmax_adj	1973	2025	7	36.0164	-76.5517	3	29.4	28.8	28.2
751	USH00315771	Tmax_adj	1973	2025	7	34.9797	-80.5233	167.6	30.8	30.7	29.1
736	USH00311458	Tmax_adj	1973	2025	7	35.2325	-75.6219	3.4	29.1	28.8	28.5
746	USH00314938	Tmax_adj	1973	2025	7	35.915	-81.5378	365.8	30	29.2	29.1
747	USH00315123	Tmax_adj	1973	2025	7	36.1028	-78.3039	79.2	31.5	31.5	30.7
757	USH00317994	Tmax_adj	1973	2025	7	35.5164	-78.3458	45.7	32.7	32.4	31.4
737	USH00311677	Tmax_adj	1973	2025	7	35.9086	-79.0794	152.4	30.4	30.9	29.7
759	USH00318292	Tmax_adj	1973	2025	7	35.81	-80.8808	289.6	29.4	29.5	28.6
748	USH00315177	Tmax_adj	1973	2025	7	34.6269	-79.025	34.1	31.5	31	30.4
739	USH00312719	Tmax_adj	1973	2025	7	36.3097	-76.205	2.4	29.8	29.8	28.8
744	USH00314055	Tmax_adj	1973	2025	7	35.0536	-83.1892	1170	23.5	22.6	22.7
755	USH00317202	Tmax_adj	1973	2025	7	36.3825	-79.6947	271.3	30.6	30.2	28.9
742	USH00313969	Tmax_adj	1973	2025	7	36.3481	-78.4119	146.3	30.7	30.5	30.2
758	USH00318113	Tmax_adj	1973	2025	7	33.9947	-78.0078	6.1	32.5	31.8	31.2
753	USH00315838	Tmax_adj	1973	2025	7	35.7308	-81.6717	353.6	30.1	29.5	29.6
754	USH00315890	Tmax_adj	1973	2025	7	36.4992	-80.6508	317.3	29.6	29.5	28.8
752	USH00315830	Tmax_adj	1973	2025	7	34.7336	-76.7358	3	28.6	27.5	27.1
760	USH00318500	Tmax_adj	1973	2025	7	35.8847	-77.5386	10.7	31.2	30.7	30.4
735	USH00310090	Tmax_adj	1973	2025	7	35.3992	-80.1994	185.9	29.8	29.9	28.6
762	USH00319147	Tmax_adj	1973	2025	7	35.4867	-82.9683	810.2	27.4	26.9	27.1
745	USH00314684	Tmax_adj	1973	2025	7	35.1967	-77.5433	7.3	31.6	30.7	30.6
756	USH00317615	Tmax_adj	1973	2025	7	35.6836	-80.4822	213.4	30.6	30	29.3
740	USH00313017	Tmax_adj	1973	2025	7	35.0583	-78.8583	29.3	31.2	30.9	30.4
741	USH00313510	Tmax_adj	1973	2025	7	35.3444	-77.9647	33.2	31.9	31.1	31
750	USH00315356	Tmax_adj	1973	2025	7	35.8036	-82.6658	609.6	29.3	29.3	29.5
743	USH00313976	Tmax_adj	1973	2025	7	35.3297	-82.4492	658.4	28.2	28.3	27.8
749	USH00315340	Tmax_adj	1973	2025	7	35.6631	-82.0292	446.8	30.4	29.8	29.7
763	USH00319476	Tmax_adj	1973	2025	7	35.6939	-77.9456	33.5	32.2	32.1	31
784	USH00327530	Tmax_adj	1973	2025	7	46.8886	-102.3192	752.9	27.3	30.6	28.7
780	USH00326155	Tmax_adj	1973	2025	7	46.3719	-102.3164	772.7	28.8	31.6	30.2
773	USH00324178	Tmax_adj	1973	2025	7	45.9925	-102.6442	816.9	28.9	32	30.6
776	USH00324958	Tmax_adj	1973	2025	7	48.7622	-98.3447	492.3	24	26.9	27.4
782	USH00326315	Tmax_adj	1973	2025	7	46.5414	-102.8692	804.4	28.1	31	30.2
764	USH00320941	Tmax_adj	1973	2025	7	48.8214	-100.4447	496.2	25.4	28.6	29.2
767	USH00322188	Tmax_adj	1973	2025	7	46.8911	-102.8111	749.8	28.9	31.7	30.2
772	USH00323621	Tmax_adj	1973	2025	7	47.9217	-97.0981	253	27	29.3	28.4
785	USH00328792	Tmax_adj	1973	2025	7	48.3706	-100.3908	451.1	26.6	30.2	30

779	USH00326015	Tmax_adj	1973	2025	7	46.6706	-100.2294	548.6	27	28.9	28.1
787	USH00329445	Tmax_adj	1973	2025	7	48.6061	-100.2911	449	25.3	28.7	29.4
783	USH00326947	Tmax_adj	1973	2025	7	48.9711	-97.2417	240.8	25.8	29.5	28.7
768	USH00322365	Tmax_adj	1973	2025	7	47.3467	-102.5869	671.8	29	32.1	30.8
771	USH00323594	Tmax_adj	1973	2025	7	48.4181	-97.4247	252.1	27.2	29.5	28.9
778	USH00325479	Tmax_adj	1973	2025	7	46.8128	-100.9097	533.4	28.8	31.4	29.8
766	USH00321871	Tmax_adj	1973	2025	7	48.9075	-103.2944	595	26.9	29.1	29.5
769	USH00323207	Tmax_adj	1973	2025	7	46.05	-100.6667	510.5	31.9	33.6	32.7
786	USH00329100	Tmax_adj	1973	2025	7	46.3233	-96.6108	291.4	28.5	30.8	29.9
777	USH00325220	Tmax_adj	1973	2025	7	46.4522	-97.6822	336.5	29	31.4	30.3
774	USH00324203	Tmax_adj	1973	2025	7	47.4389	-97.0664	277.4	28.7	30.9	30.7
781	USH00326255	Tmax_adj	1973	2025	7	46.5067	-99.7692	603.5	28.3	31.7	30.4
775	USH00324418	Tmax_adj	1973	2025	7	46.8844	-98.685	447.1	29.7	32.2	30.2
765	USH00321408	Tmax_adj	1973	2025	7	46.8769	-97.2328	285	28.5	30.4	30.2
770	USH00323287	Tmax_adj	1973	2025	7	46.1581	-98.4	437.4	30.4	32.2	30
580	USH00251145	Tmax_adj	1973	2025	7	41.6675	-103.1042	1117	31.1	33.5	32.6
576	USH00250420	Tmax_adj	1973	2025	7	42.5136	-99.0303	637	29.4	34.3	30.9
598	USH00254440	Tmax_adj	1973	2025	7	41.2453	-103.6344	1435	28.3	31.8	31.1
608	USH00256970	Tmax_adj	1973	2025	7	42.065	-100.2472	819.9	30.5	34.4	31.4
599	USH00254900	Tmax_adj	1973	2025	7	41.1489	-102.6361	1168	30.9	33.8	32
574	USH00250130	Tmax_adj	1973	2025	7	42.1103	-102.8967	1217	29.6	32.8	31.9
594	USH00253715	Tmax_adj	1973	2025	7	42.5119	-102.6944	1160	30.1	34.3	32.6
609	USH00257070	Tmax_adj	1973	2025	7	40.0978	-98.5197	524.3	31.6	35.7	32.3
591	USH00253615	Tmax_adj	1973	2025	7	42.6858	-103.8842	1478	28.8	31.9	31.2
597	USH00254110	Tmax_adj	1973	2025	7	40.5208	-101.655	999.7	31.4	34.5	32.8
603	USH00255470	Tmax_adj	1973	2025	7	42.9194	-101.7089	986	31.6	35.2	32.3
602	USH00255310	Tmax_adj	1973	2025	7	40.2156	-100.6203	796.1	31.7	36.3	33.8
610	USH00257515	Tmax_adj	1973	2025	7	41.2686	-98.4697	541	30.3	35.1	31.1
575	USH00250375	Tmax_adj	1973	2025	7	41.0406	-96.3775	326.1	29.4	36	31.8
604	USH00255565	Tmax_adj	1973	2025	7	40.5156	-98.9514	658.4	29.6	34.3	30.8
605	USH00256040	Tmax_adj	1973	2025	7	41.4933	-98.7747	597.4	29.6	34	30.5
615	USH00258480	Tmax_adj	1973	2025	7	41.7861	-96.2264	338.3	30.6	35.6	32.5
581	USH00251200	Tmax_adj	1973	2025	7	41.4083	-99.675	762	29.8	33.7	30.6
596	USH00253910	Tmax_adj	1973	2025	7	40.4517	-99.3803	707.1	30.6	34.4	30.9
612	USH00258133	Tmax_adj	1973	2025	7	41.4581	-100.5986	911.4	30.6	35.5	31.2
587	USH00253035	Tmax_adj	1973	2025	7	40.1	-98.9667	565.4	31.2	35.6	32.5
582	USH00252020	Tmax_adj	1973	2025	7	40.6194	-96.9469	437.4	29.7	35.7	31.5
616	USH00258915	Tmax_adj	1973	2025	7	42.2667	-96.8617	423.7	28.9	34.4	31.2
600	USH00254985	Tmax_adj	1973	2025	7	41.2808	-98.9681	627.3	29.8	34.8	32.4
578	USH00250622	Tmax_adj	1973	2025	7	40.2994	-96.75	395.3	30.2	35.5	31.2
585	USH00252820	Tmax_adj	1973	2025	7	40.0739	-97.1669	411.5	31	37.1	32.7
590	USH00253365	Tmax_adj	1973	2025	7	40.94	-100.1522	787.9	30.4	33.8	31.1
583	USH00252100	Tmax_adj	1973	2025	7	40.6742	-100.4936	829.4	32.6	37.2	33.3
586	USH00252840	Tmax_adj	1973	2025	7	40.6422	-97.5928	499.9	30.2	35.8	31.2
584	USH00252205	Tmax_adj	1973	2025	7	41.2492	-97.1328	490.7	30.1	35	31.3
618	USH00259510	Tmax_adj	1973	2025	7	40.8678	-97.5922	490.7	32.2	37.4	33.4
579	USH00250640	Tmax_adj	1973	2025	7	40.1306	-99.8278	658.4	33.4	37.4	35.1

589	USH00253185	Tmax_adj	1973	2025	7	41.4514	-97.7644	484.6	29.5	35.6	31.3
601	USH00255080	Tmax_adj	1973	2025	7	41.8292	-97.45	481.6	30	34.5	31.5
617	USH00259090	Tmax_adj	1973	2025	7	40.8639	-96.1414	335.3	28.9	35.1	31.2
588	USH00253175	Tmax_adj	1973	2025	7	40.5314	-97.5964	496.8	30.3	35.5	31.1
595	USH00253735	Tmax_adj	1973	2025	7	40.175	-97.5903	451.1	31.1	36.1	32.4
607	USH00256570	Tmax_adj	1973	2025	7	40.1217	-96.1558	378	30.3	36.1	31.6
606	USH00256135	Tmax_adj	1973	2025	7	42.0678	-97.9675	521.2	30.7	34.2	30.6
592	USH00253630	Tmax_adj	1973	2025	7	42.6167	-97.2608	417.6	30.8	35.2	32.9
593	USH00253660	Tmax_adj	1973	2025	7	40.6475	-98.3833	591.3	30.9	35.6	32.1
613	USH00258395	Tmax_adj	1973	2025	7	40.6825	-96.1886	335.3	31.2	37	33
611	USH00257715	Tmax_adj	1973	2025	7	40.9	-97.0908	438.9	31.9	38	33.5
614	USH00258465	Tmax_adj	1973	2025	7	40.3544	-96.1939	338.3	30.7	36.6	33.2
577	USH00250435	Tmax_adj	1973	2025	7	40.3703	-95.7467	283.5	30.6	35.8	32
573	USH00250070	Tmax_adj	1973	2025	7	41.6867	-98.0036	545.6	30.2	35.6	31.7
636	USH00274399	Tmax_adj	1973	2025	7	42.9389	-72.3244	158.5	26.7	26.7	28.2
633	USH00272174	Tmax_adj	1973	2025	7	43.15	-70.95	24.4	27.1	25.4	27.6
634	USH00272999	Tmax_adj	1973	2025	7	45.0875	-71.2872	506	24.5	22.9	25.5
632	USH00270706	Tmax_adj	1973	2025	7	44.3061	-71.6575	359.7	26.9	25.4	27.4
635	USH00273850	Tmax_adj	1973	2025	7	43.7031	-72.2847	183.8	29.3	27.9	30.1
638	USH00280734	Tmax_adj	1973	2025	7	40.8292	-75.0836	80.2	27.8	28.2	27.5
642	USH00283951	Tmax_adj	1973	2025	7	40.265	-74.5642	30.5	29.6	29.5	29.3
639	USH00280907	Tmax_adj	1973	2025	7	40.9	-74.4	85.3	26.9	28.5	28.1
645	USH00285728	Tmax_adj	1973	2025	7	39.9511	-74.9697	13.7	28.3	28.6	26.9
646	USH00286055	Tmax_adj	1973	2025	7	40.4719	-74.4364	26.2	29.7	29.8	29.2
644	USH00284987	Tmax_adj	1973	2025	7	40.2797	-74.0047	9.1	28.3	27.9	27.5
641	USH00283029	Tmax_adj	1973	2025	7	40.5631	-74.8831	79.2	29.5	29.4	28.6
647	USH00287079	Tmax_adj	1973	2025	7	40.6036	-74.4025	27.4	30.4	30.1	30.2
648	USH00288816	Tmax_adj	1973	2025	7	39.95	-74.2167	30.5	28.3	29.5	28
637	USH00280325	Tmax_adj	1973	2025	7	39.3792	-74.4242	3	25.8	27.3	25.4
640	USH00281582	Tmax_adj	1973	2025	7	41.0347	-74.4233	231.6	28.2	27.6	27.5
643	USH00284229	Tmax_adj	1973	2025	7	39.8144	-74.7883	30.5	30.6	30.2	29.4
664	USH00295150	Tmax_adj	1973	2025	7	34.7675	-106.7611	1475	31.1	31.5	31.5
670	USH00297610	Tmax_adj	1973	2025	7	33.3075	-104.5083	1112	33.3	34.2	31.5
649	USH00290692	Tmax_adj	1973	2025	7	36.835	-108.0006	1720	31.9	31.5	31.7
673	USH00298387	Tmax_adj	1973	2025	7	34.0828	-106.8831	1398	31.2	31.1	31.2
676	USH00299156	Tmax_adj	1973	2025	7	35.2006	-103.6867	1245	32.3	34	30.9
669	USH00297323	Tmax_adj	1973	2025	7	36.7058	-105.4036	2644	21.3	21.9	20.2
677	USH00299165	Tmax_adj	1973	2025	7	33.0719	-106.0417	1350	30.8	31.3	31
660	USH00293368	Tmax_adj	1973	2025	7	32.2261	-108.0867	1365	32.7	32.4	32.5
666	USH00295960	Tmax_adj	1973	2025	7	32.9547	-105.8242	2067	23	23.8	22.4
662	USH00294426	Tmax_adj	1973	2025	7	32.6169	-106.7411	1300	33.2	32.5	32.5
652	USH00291515	Tmax_adj	1973	2025	7	33.6308	-105.8964	1647	30.1	30.4	28.7
665	USH00295273	Tmax_adj	1973	2025	7	33.8225	-108.9417	2149	27.5	27.4	27.6
650	USH00290858	Tmax_adj	1973	2025	7	35.5297	-104.0936	1372	32.4	34	31
658	USH00293265	Tmax_adj	1973	2025	7	32.7939	-108.1517	1872	30.2	29.6	29.6
657	USH00292848	Tmax_adj	1973	2025	7	33.1461	-107.1844	1395	33.7	32.3	32.8
671	USH00297867	Tmax_adj	1973	2025	7	35.1086	-103.3283	1289	33	35	31.2

659	USH00293294	Tmax_adj	1973	2025	7	34.4667	-104.2319	1227	31.7	32.7	29.6
656	USH00292608	Tmax_adj	1973	2025	7	36.9358	-107	2071	29	28.5	28.5
667	USH00295965	Tmax_adj	1973	2025	7	34.5208	-106.2606	1987	27.9	27.1	27.6
674	USH00298501	Tmax_adj	1973	2025	7	36.3628	-104.585	1795	31	31.9	30
663	USH00294862	Tmax_adj	1973	2025	7	35.5678	-105.2131	1935	27.1	28.2	26.8
675	USH00298535	Tmax_adj	1973	2025	7	32.2822	-106.7597	1183	33.1	32.7	33.2
672	USH00298107	Tmax_adj	1973	2025	7	34.9358	-104.6805	1405	32.2	32.4	30.5
655	USH00291887	Tmax_adj	1973	2025	7	36.4486	-103.1539	1512	30.2	31.3	29.1
651	USH00291469	Tmax_adj	1973	2025	7	32.3478	-104.2225	951	34.4	35.7	32.6
653	USH00291664	Tmax_adj	1973	2025	7	36.9178	-106.5781	2393	25.4	25.4	24.5
654	USH00291813	Tmax_adj	1973	2025	7	36.4661	-104.9456	1993	26.9	29.2	25.8
668	USH00296435	Tmax_adj	1973	2025	7	32.3789	-106.0911	1275	35.7	33.6	33
661	USH00294369	Tmax_adj	1973	2025	7	35.7783	-106.6872	1909	30.2	30.5	29.5
623	USH00262780	Tmax_adj	1973	2025	7	39.4572	-118.7811	1209	32.8	31.9	32.7
622	USH00262573	Tmax_adj	1973	2025	7	40.8289	-115.7886	1533	32.2	31.8	32.3
625	USH00264698	Tmax_adj	1973	2025	7	40.1906	-118.4767	1212	32.9	32.1	32.8
628	USH00266779	Tmax_adj	1973	2025	7	39.4839	-119.7711	1344	33.8	31	33.4
631	USH00269171	Tmax_adj	1973	2025	7	40.9017	-117.8081	1309	34.4	33.5	33.9
627	USH00265168	Tmax_adj	1973	2025	7	38.3844	-118.1056	1392	35.8	35	35.6
626	USH00264950	Tmax_adj	1973	2025	7	39.4136	-114.7733	1911	29.6	29.8	30.1
620	USH00260691	Tmax_adj	1973	2025	7	40.6117	-116.8917	1373	36.3	35.6	37
630	USH00268988	Tmax_adj	1973	2025	7	41.1006	-114.9736	1737	30.2	30.9	30.8
624	USH00263245	Tmax_adj	1973	2025	7	40.9567	-117.4922	1339	33.1	32.8	33.4
619	USH00260507	Tmax_adj	1973	2025	7	39.4931	-117.0675	2067	29.2	28.1	27.9
629	USH00267369	Tmax_adj	1973	2025	7	35.4661	-114.9217	1079	36.7	35	36
621	USH00261071	Tmax_adj	1973	2025	7	35.98	-114.8464	762	38.7	36.9	38.7
713	USH00305426	Tmax_adj	1973	2025	7	41.7681	-74.155	379.5	26.3	26	25.6
705	USH00304555	Tmax_adj	1973	2025	7	44.2444	-73.9847	591.3	22.8	20.8	24
700	USH00303259	Tmax_adj	1973	2025	7	41.5167	-73.9333	83.8	28.6	27.5	28.4
686	USH00300889	Tmax_adj	1973	2025	7	40.9464	-72.3067	18.3	25.5	26	25.4
719	USH00306774	Tmax_adj	1973	2025	7	41.38	-74.6847	143.3	28.1	27.9	27.8
697	USH00302610	Tmax_adj	1973	2025	7	42.0997	-76.8358	288.6	28.4	28.2	28.4
714	USH00305512	Tmax_adj	1973	2025	7	42.8417	-75.7264	396.2	25.3	24.4	25.6
704	USH00304174	Tmax_adj	1973	2025	7	42.4489	-76.4489	292.6	26.9	26.2	26.9
715	USH00305801	Tmax_adj	1973	2025	7	40.7789	-73.9692	39.6	28.6	29.3	27.6
733	USH00309292	Tmax_adj	1973	2025	7	41.3906	-73.9608	97.5	28.7	28.1	28.1
679	USH00300042	Tmax_adj	1973	2025	7	42.7431	-73.8092	83.8	29.7	27.6	28.7
692	USH00301799	Tmax_adj	1973	2025	7	42.6	-76.1833	344.1	28	26.7	27.6
723	USH00307633	Tmax_adj	1973	2025	7	40.9586	-73.1047	12.2	28.5	29	27.9
678	USH00300023	Tmax_adj	1973	2025	7	42.1014	-77.2344	304.5	28.6	27.6	28.2
680	USH00300085	Tmax_adj	1973	2025	7	42.2614	-77.785	539.5	27.2	26.6	27.3
726	USH00308600	Tmax_adj	1973	2025	7	42.75	-73.6831	7.3	30	29	30.2
685	USH00300687	Tmax_adj	1973	2025	7	42.2067	-75.98	486.2	26.8	26.2	28.1
691	USH00301752	Tmax_adj	1973	2025	7	42.7167	-74.9267	383.1	25.8	25	26.1
690	USH00301401	Tmax_adj	1973	2025	7	44.8786	-73.3953	47.9	28.2	26.9	28.5
696	USH00302129	Tmax_adj	1973	2025	7	41.0072	-73.8344	61	29.1	29.5	28.4
725	USH00308383	Tmax_adj	1973	2025	7	43.1092	-76.1033	125	29.2	27.1	28.5

709	USH00304844	Tmax_adj	1973	2025	7	43.1392	-78.6814	184.4	26.8	26	26.6
716	USH00306085	Tmax_adj	1973	2025	7	42.5117	-75.5197	301.4	27	26.4	27.3
720	USH00306820	Tmax_adj	1973	2025	7	41.5997	-73.91	51.8	28.6	27.8	28.6
734	USH00309670	Tmax_adj	1973	2025	7	41.2664	-73.7975	204.2	28.6	28.5	28.4
710	USH00304912	Tmax_adj	1973	2025	7	43.7975	-75.4817	262.1	26.1	24.8	26.8
706	USH00304647	Tmax_adj	1973	2025	7	44.7583	-74.6692	142	26.6	25.1	27.7
728	USH00308737	Tmax_adj	1973	2025	7	43.145	-75.3839	216.7	27.7	26	27.1
699	USH00303184	Tmax_adj	1973	2025	7	42.8767	-77.0308	218.8	26.6	25.8	26.9
727	USH00308631	Tmax_adj	1973	2025	7	44.2308	-74.4383	512.1	26.2	24.4	26.9
684	USH00300443	Tmax_adj	1973	2025	7	43.0303	-78.1692	278.3	26.7	26	26.7
695	USH00302060	Tmax_adj	1973	2025	7	42.0628	-75.4264	304.8	28.4	27.2	27.9
707	USH00304791	Tmax_adj	1973	2025	7	43.0603	-74.8686	274.3	29.4	27.1	27.9
689	USH00301185	Tmax_adj	1973	2025	7	44.5772	-75.1097	136.6	27.2	25.8	28.3
731	USH00308944	Tmax_adj	1973	2025	7	44.1481	-74.9003	460.2	26.4	25.4	27.1
729	USH00308906	Tmax_adj	1973	2025	7	41.5514	-74.1628	115.8	29.3	29.2	29.4
681	USH00300093	Tmax_adj	1973	2025	7	42.1003	-78.75	457.2	26.5	26	26.9
694	USH00301974	Tmax_adj	1973	2025	7	42.5656	-77.7175	201.2	29.2	28.6	29.4
708	USH00304796	Tmax_adj	1973	2025	7	43.0333	-74.8667	109.7	29.4	27.6	29.4
698	USH00303033	Tmax_adj	1973	2025	7	42.4497	-79.312	231.6	27.1	26.4	26.5
687	USH00300937	Tmax_adj	1973	2025	7	43.2	-77.9333	163.1	27.6	27.3	28
703	USH00304102	Tmax_adj	1973	2025	7	43.755	-74.2692	506	24.8	23.5	25.8
682	USH00300183	Tmax_adj	1973	2025	7	42.3017	-77.9889	440.4	28.3	27.6	28.2
688	USH00301012	Tmax_adj	1973	2025	7	42.9408	-78.7358	214.9	27.3	26.3	27.7
701	USH00303319	Tmax_adj	1973	2025	7	43.0492	-74.3592	246.9	27.4	27	28.1
732	USH00309000	Tmax_adj	1973	2025	7	43.9761	-75.8753	151.5	27.9	24.8	27.2
712	USH00305113	Tmax_adj	1973	2025	7	42.4694	-75.0106	373.4	26.3	25.2	26.1
717	USH00306164	Tmax_adj	1973	2025	7	44.7281	-75.4442	85.3	27.3	26	29
683	USH00300321	Tmax_adj	1973	2025	7	42.9328	-76.5447	234.7	27.7	26.5	28.1
724	USH00308248	Tmax_adj	1973	2025	7	42.6925	-73.8325	515.1	25.2	23.5	25.5
722	USH00307484	Tmax_adj	1973	2025	7	43.0331	-73.8167	94.5	28.9	27.9	28.8
711	USH00304996	Tmax_adj	1973	2025	7	44.8419	-74.3081	268.2	26.8	24.6	28.3
721	USH00307167	Tmax_adj	1973	2025	7	43.1167	-77.6767	162.5	28.5	27.8	28.6
693	USH00301966	Tmax_adj	1973	2025	7	44.7192	-73.7206	408.4	25.8	24.7	28.1
730	USH00308910	Tmax_adj	1973	2025	7	42.7411	-78.51	332.2	27.8	27.4	28
702	USH00303773	Tmax_adj	1973	2025	7	42.7747	-77.6083	274.9	28.4	27.5	28.1
718	USH00306314	Tmax_adj	1973	2025	7	43.4622	-76.4933	106.7	27.5	26.1	27.4
802	USH00335315	Tmax_adj	1973	2025	7	40.7692	-80.8561	356.6	27.1	27.8	27.5
810	USH00338769	Tmax_adj	1973	2025	7	41.2	-80.8167	274.3	26.2	27	27.7
789	USH00331152	Tmax_adj	1973	2025	7	40.2686	-80.9981	384	27.1	27	27
798	USH00333780	Tmax_adj	1973	2025	7	41.3	-81.15	374.9	26.1	26.4	26.5
790	USH00331541	Tmax_adj	1973	2025	7	41.0517	-81.9361	359.7	26.4	27.5	27.5
813	USH00339312	Tmax_adj	1973	2025	7	40.7833	-81.9167	310.9	27.8	28.9	29
806	USH00336781	Tmax_adj	1973	2025	7	38.7569	-82.8872	164.6	30.3	30	29.6
792	USH00331890	Tmax_adj	1973	2025	7	40.2403	-81.8711	231.6	28	28.6	28.7
797	USH00333758	Tmax_adj	1973	2025	7	39.2	-83.6167	335.3	28.4	28.7	28.3
801	USH00335297	Tmax_adj	1973	2025	7	40.55	-81.9167	249.6	27.9	29.3	28.4
793	USH00332098	Tmax_adj	1973	2025	7	41.2778	-84.3853	213.4	28.3	30.5	28.4

812	USH00338830	Tmax_adj	1973	2025	7	39.1114	-82.9797	170.7	29.3	30.3	29.9
803	USH00336118	Tmax_adj	1973	2025	7	41.2667	-82.6167	204.2	28.4	28.9	28.6
807	USH00338313	Tmax_adj	1973	2025	7	41.1167	-83.1667	225.6	27.9	29.6	28.3
804	USH00336196	Tmax_adj	1973	2025	7	41.2667	-82.2167	248.7	28.5	29.8	29.1
805	USH00336600	Tmax_adj	1973	2025	7	39.8386	-81.9167	310.9	28.6	29.4	29.9
809	USH00338552	Tmax_adj	1973	2025	7	40.1	-83.7833	304.8	28.3	30.2	28.5
799	USH00334189	Tmax_adj	1973	2025	7	40.6489	-83.6061	303.3	28.5	30.3	28.4
794	USH00332119	Tmax_adj	1973	2025	7	40.3175	-83.0739	280.4	29.6	30.8	29.9
791	USH00331592	Tmax_adj	1973	2025	7	39.6106	-82.9547	205.1	28.8	29.6	29.5
800	USH00335041	Tmax_adj	1973	2025	7	39.6539	-81.8569	231.6	29	29.7	29.3
811	USH00338822	Tmax_adj	1973	2025	7	41.5183	-84.1453	228.6	28.5	30.9	28.6
788	USH00331072	Tmax_adj	1973	2025	7	40.8128	-82.9694	291.1	29.1	31.5	29.6
795	USH00332791	Tmax_adj	1973	2025	7	41.0461	-83.6622	234.1	29.1	30.8	29.3
808	USH00338534	Tmax_adj	1973	2025	7	40.8333	-83.2833	260.3	29.3	31	29.8
796	USH00333375	Tmax_adj	1973	2025	7	40.1	-84.65	312.1	29.1	30.4	29.4
833	USH00344204	Tmax_adj	1973	2025	7	34.9894	-99.0525	474.3	33.2	36.6	31.4
837	USH00344766	Tmax_adj	1973	2025	7	36.9031	-102.965	1326	31.2	33	31.6
829	USH00343628	Tmax_adj	1973	2025	7	36.5914	-101.6181	1009	31.6	34.4	31.4
820	USH00340908	Tmax_adj	1973	2025	7	36.7236	-102.4806	1260	31.1	33.4	31.2
822	USH00341504	Tmax_adj	1973	2025	7	35.1756	-98.5794	451.4	33.2	36.6	31.7
815	USH00340179	Tmax_adj	1973	2025	7	34.5903	-99.3344	420.6	35.1	37.2	32.4
816	USH00340256	Tmax_adj	1973	2025	7	34.2208	-95.615	143.3	32.2	34.9	32.7
827	USH00342944	Tmax_adj	1973	2025	7	35.2164	-99.8628	627.9	34.1	36.3	31.7
857	USH00349445	Tmax_adj	1973	2025	7	35.4814	-95.2039	167.6	34.2	37	34.1
828	USH00343497	Tmax_adj	1973	2025	7	35.6267	-98.3225	487.7	32.7	36.2	31.8
834	USH00344235	Tmax_adj	1973	2025	7	35.0567	-96.3861	260.6	33.4	36.6	32.9
855	USH00349395	Tmax_adj	1973	2025	7	34.1747	-97.9964	278	33.6	37.4	32.5
814	USH00340017	Tmax_adj	1973	2025	7	34.7864	-96.685	309.4	32.8	35.6	32.8
846	USH00346629	Tmax_adj	1973	2025	7	36.1217	-98.315	370.3	33.6	36.9	33.1
830	USH00343821	Tmax_adj	1973	2025	7	35.8161	-97.395	338.3	32.7	35.7	32
849	USH00346926	Tmax_adj	1973	2025	7	34.7253	-97.2814	286.5	33.3	35.8	32.5
817	USH00340292	Tmax_adj	1973	2025	7	34.1714	-97.1294	268.2	33.1	35.8	32.6
852	USH00347254	Tmax_adj	1973	2025	7	35.0539	-94.6264	134.1	32.6	35.6	32.4
847	USH00346638	Tmax_adj	1973	2025	7	35.4253	-96.3033	285	32.9	35.9	32.8
832	USH00344055	Tmax_adj	1973	2025	7	36.0942	-97.835	357.8	34.1	36.8	33.2
835	USH00344298	Tmax_adj	1973	2025	7	36.8589	-101.2172	912.9	33.6	36	33.5
851	USH00347012	Tmax_adj	1973	2025	7	36.2886	-97.2897	312.4	34.1	36.5	33.6
841	USH00345779	Tmax_adj	1973	2025	7	35.505	-96.9767	281.9	32.1	35.1	31.5
848	USH00346670	Tmax_adj	1973	2025	7	35.6239	-96.025	197.2	34	35.9	32.7
856	USH00349422	Tmax_adj	1973	2025	7	35.52	-98.6986	493.2	34.1	36.6	32.4
819	USH00340593	Tmax_adj	1973	2025	7	36.8125	-100.5308	751.3	33.9	35.8	33.1
821	USH00341243	Tmax_adj	1973	2025	7	36.8003	-99.6403	588.3	34.5	36.5	33.3
839	USH00345063	Tmax_adj	1973	2025	7	34.6097	-98.4572	350.5	34.5	37.2	32.5
853	USH00348501	Tmax_adj	1973	2025	7	36.1175	-97.095	272.8	33.8	36.5	32.9
823	USH00341724	Tmax_adj	1973	2025	7	36.7747	-98.3583	359.7	34.9	37.4	33.5
826	USH00342912	Tmax_adj	1973	2025	7	36.4194	-97.8747	379.5	33.6	36.8	33.3
843	USH00346130	Tmax_adj	1973	2025	7	35.7781	-95.3339	157.9	32.1	35.7	32.7

825	USH00342678	Tmax_adj	1973	2025	7	34.0003	-96.3686	182.9	33.5	36.7	34.1
831	USH00343871	Tmax_adj	1973	2025	7	35.585	-99.3953	554.7	34.3	37.3	33.1
840	USH00345509	Tmax_adj	1973	2025	7	34.8911	-99.5017	486.2	34.1	37.2	31.7
836	USH00344573	Tmax_adj	1973	2025	7	36.7222	-97.7903	318.5	34.4	37.2	33.7
838	USH00344861	Tmax_adj	1973	2025	7	35.8583	-97.9294	320	34	37.4	33.4
818	USH00340548	Tmax_adj	1973	2025	7	36.7683	-96.0261	217.9	33.8	36.2	33.5
854	USH00348677	Tmax_adj	1973	2025	7	35.9369	-94.9644	259.1	32.8	36.3	32.7
824	USH00341828	Tmax_adj	1973	2025	7	36.3225	-95.5808	179.2	32.4	36.4	33.5
842	USH00345855	Tmax_adj	1973	2025	7	36.8833	-94.8833	245.4	33.5	36.8	33.3
850	USH00346935	Tmax_adj	1973	2025	7	36.6692	-96.3472	254.5	32.7	36.2	32.8
844	USH00346139	Tmax_adj	1973	2025	7	36.2283	-99.17	576.1	34.5	36.8	32.6
845	USH00346278	Tmax_adj	1973	2025	7	36.8914	-97.0586	347.5	33.3	37.2	32.7
861	USH00350694	Tmax_adj	1973	2025	7	44.0569	-121.285	1116	27.3	25.8	27.1
873	USH00353445	Tmax_adj	1973	2025	7	42.4244	-123.3236	283.5	32.9	30.1	30.7
881	USH00355362	Tmax_adj	1973	2025	7	44.1781	-122.1156	450.5	28.7	26.5	26.9
895	USH00358746	Tmax_adj	1973	2025	7	45.2081	-117.8758	842.8	30.1	27.2	28.7
880	USH00355162	Tmax_adj	1973	2025	7	43.265	-118.8447	1255	27.3	25.5	26.8
887	USH00356426	Tmax_adj	1973	2025	7	42.6922	-120.5403	1329	28.8	27	27.4
892	USH00357331	Tmax_adj	1973	2025	7	43.2131	-123.3658	129.5	31.4	29.6	29.8
870	USH00352440	Tmax_adj	1973	2025	7	45.4539	-121.1303	405.4	30.4	28	30
877	USH00354003	Tmax_adj	1973	2025	7	45.6847	-121.5175	152.4	26.3	24.3	25
868	USH00352135	Tmax_adj	1973	2025	7	42.9447	-117.3389	1288	29.6	28.4	29.3
884	USH00355734	Tmax_adj	1973	2025	7	45.4825	-120.7236	570	28.7	26.1	28.6
858	USH00350304	Tmax_adj	1973	2025	7	42.2128	-122.7144	532.2	31.8	28.9	28.9
896	USH00358797	Tmax_adj	1973	2025	7	43.9814	-117.2439	682.8	33.2	31.2	32.4
897	USH00358997	Tmax_adj	1973	2025	7	45.5722	-117.5314	890.9	31	28.8	29.7
867	USH00351946	Tmax_adj	1973	2025	7	42.8967	-122.1328	1974	21.4	17.6	19.3
869	USH00352406	Tmax_adj	1973	2025	7	43.6656	-123.3275	89	29.4	25.9	27.5
860	USH00350412	Tmax_adj	1973	2025	7	44.8428	-117.8086	1024	30.3	27.7	29.1
863	USH00351433	Tmax_adj	1973	2025	7	44.3914	-122.4811	292.6	25.6	23.3	24.2
865	USH00351862	Tmax_adj	1973	2025	7	44.6342	-123.19	68.6	28.5	25.5	26.8
879	USH00354670	Tmax_adj	1973	2025	7	42.2139	-120.3636	1456	30	27.5	29.5
864	USH00351765	Tmax_adj	1973	2025	7	45.2325	-120.1817	865.6	29.6	27.1	29.4
878	USH00354506	Tmax_adj	1973	2025	7	42.2008	-121.7814	1249	32	29.3	30.4
882	USH00355384	Tmax_adj	1973	2025	7	45.2214	-123.1622	47.2	29	25.5	26.3
883	USH00355593	Tmax_adj	1973	2025	7	45.9428	-118.4089	295.7	31.1	29	31.5
890	USH00356907	Tmax_adj	1973	2025	7	42.7342	-122.5164	756.5	30.8	28.2	28.7
876	USH00353847	Tmax_adj	1973	2025	7	45.8289	-119.2636	195.1	32.5	30.7	33.1
893	USH00358466	Tmax_adj	1973	2025	7	45.1219	-122.07	341.4	25.7	23.2	24.4
866	USH00351897	Tmax_adj	1973	2025	7	43.7917	-123.0275	181.4	28	25.7	25.8
872	USH00353095	Tmax_adj	1973	2025	7	43.3936	-121.2122	1405	30.3	28.1	29.6
891	USH00357169	Tmax_adj	1973	2025	7	42.9506	-123.3572	207.3	31.1	29.1	29.5
888	USH00356634	Tmax_adj	1973	2025	7	45.4756	-118.8253	524.3	32.7	30.6	32.6
874	USH00353770	Tmax_adj	1973	2025	7	45.4486	-122.1547	228	24.9	22	22.8
889	USH00356883	Tmax_adj	1973	2025	7	44.3022	-120.8081	888.5	31	28.6	31.1
871	USH00352997	Tmax_adj	1973	2025	7	45.5247	-123.1028	54.9	28.2	25.6	26.8
875	USH00353827	Tmax_adj	1973	2025	7	45.3653	-119.5639	574.5	31.2	28.9	30.9

862	USH00351055	Tmax_adj	1973	2025	7	42.03	-124.2453	15.2	16.8	16.8	17.3
859	USH00350328	Tmax_adj	1973	2025	7	46.1569	-123.8825	2.7	19.8	19.2	19.1
894	USH00358494	Tmax_adj	1973	2025	7	45.4567	-123.8728	3	19.4	19	18.4
886	USH00356073	Tmax_adj	1973	2025	7	43.4133	-124.2436	1.8	18.9	17.9	18.7
885	USH00356032	Tmax_adj	1973	2025	7	44.6431	-124.0556	37.2	18.1	18.3	17.1
918	USH00369408	Tmax_adj	1973	2025	7	41.7003	-77.3894	554.1	24.8	24.3	24.3
909	USH00367029	Tmax_adj	1973	2025	7	41.7394	-75.4464	548.6	24.3	23.4	23.5
920	USH00369728	Tmax_adj	1973	2025	7	41.2433	-76.9217	158.5	28	28.2	28
907	USH00366233	Tmax_adj	1973	2025	7	41.0172	-80.3619	251.5	29.2	29	28.8
898	USH00360106	Tmax_adj	1973	2025	7	40.6508	-75.4492	118.9	28.9	28.1	27.8
910	USH00367322	Tmax_adj	1973	2025	7	40.4269	-75.9319	109.7	29.6	29.6	29.5
912	USH00367931	Tmax_adj	1973	2025	7	40.7831	-76.8611	128	29.2	28.2	28.4
919	USH00369464	Tmax_adj	1973	2025	7	39.9708	-75.635	114.3	28.9	28.9	28.3
906	USH00365915	Tmax_adj	1973	2025	7	41.8511	-75.8583	432.8	26.4	25.6	26.2
900	USH00362537	Tmax_adj	1973	2025	7	39.805	-77.2292	164.6	29.2	29.1	28.8
905	USH00364896	Tmax_adj	1973	2025	7	40.3333	-76.4667	137.2	27.7	28.8	27.5
899	USH00361354	Tmax_adj	1973	2025	7	39.9353	-77.6394	195.1	29.4	29.7	29.2
915	USH00368905	Tmax_adj	1973	2025	7	41.7511	-76.4431	231.6	28.3	28	28.3
921	USH00369933	Tmax_adj	1973	2025	7	39.9167	-76.75	118.9	30.3	29.9	29.4
911	USH00367477	Tmax_adj	1973	2025	7	41.42	-78.7497	414.5	26.9	27.1	28.1
902	USH00363028	Tmax_adj	1973	2025	7	41.4003	-79.8303	309.4	28	27.6	28
908	USH00366689	Tmax_adj	1973	2025	7	40.8	-75.6167	125	32.2	31.4	32
903	USH00363526	Tmax_adj	1973	2025	7	41.4167	-80.3667	344.4	27.6	28.2	28.1
913	USH00368449	Tmax_adj	1973	2025	7	40.7933	-77.8672	356.6	27.5	27	27.9
901	USH00362682	Tmax_adj	1973	2025	7	42.08	-80.1825	222.5	26.6	25.2	27.5
904	USH00364385	Tmax_adj	1973	2025	7	40.3333	-78.9167	370	30.5	31	31.2
916	USH00369050	Tmax_adj	1973	2025	7	39.915	-79.7192	291.4	28.8	28.4	29.4
917	USH00369298	Tmax_adj	1973	2025	7	41.85	-79.15	368.8	27.6	27.3	28.2
914	USH00368596	Tmax_adj	1973	2025	7	41.0125	-75.1906	140.2	29.1	29.4	28.9
923	USH00374266	Tmax_adj	1973	2025	7	41.4906	-71.5414	34.7	27	26.7	26.6
922	USH00370896	Tmax_adj	1973	2025	7	41.1667	-71.5833	33.5	25.5	25.1	25.8
924	USH00376698	Tmax_adj	1973	2025	7	41.7219	-71.4325	15.5	27.3	27	27.4
934	USH00381944	Tmax_adj	1973	2025	7	33.9831	-81.0167	73.8	32.2	31.5	30.4
953	USH00389469	Tmax_adj	1973	2025	7	32.68	-80.8417	7.6	32.3	30.9	30
947	USH00387722	Tmax_adj	1973	2025	7	34.635	-81.5206	158.5	31.5	31.3	29.8
950	USH00388887	Tmax_adj	1973	2025	7	34.7483	-83.0842	298.7	29.9	29.7	29.4
951	USH00389327	Tmax_adj	1973	2025	7	34.3739	-81.0928	170.7	30.9	29.8	28.5
933	USH00381770	Tmax_adj	1973	2025	7	34.6603	-82.8236	251.2	30.8	30.6	30.9
936	USH00382260	Tmax_adj	1973	2025	7	34.3011	-79.8767	45.7	32.1	31.6	30.2
928	USH00380764	Tmax_adj	1973	2025	7	33.3631	-81.3292	98.8	32.4	31.2	30.5
926	USH00380165	Tmax_adj	1973	2025	7	34.5283	-82.6606	243.8	31.5	31.1	30.2
929	USH00381277	Tmax_adj	1973	2025	7	34.0906	-82.5883	161.5	30.9	31.3	30.5
931	USH00381549	Tmax_adj	1973	2025	7	32.78	-79.9319	3	29.7	28.5	29.4
948	USH00388426	Tmax_adj	1973	2025	7	33.0367	-80.2325	19.8	32.4	31	30.5
942	USH00385017	Tmax_adj	1973	2025	7	34.4989	-82.0219	179.5	31.3	30.2	29.9
932	USH00381588	Tmax_adj	1973	2025	7	34.7319	-79.8833	42.7	31.5	30.8	29.7
930	USH00381310	Tmax_adj	1973	2025	7	34.2428	-80.6564	42.7	31.4	30.4	29.7

943	USH00385200	Tmax_adj	1973	2025	7	34.1942	-81.415	216.7	31.4	31.7	29.8
944	USH00386209	Tmax_adj	1973	2025	7	34.2978	-81.6236	145.1	32.3	31.4	30.3
927	USH00380559	Tmax_adj	1973	2025	7	32.3953	-80.6956	7.6	32.5	31.9	30.7
935	USH00381997	Tmax_adj	1973	2025	7	33.8314	-79.0558	6.1	32.7	31.8	31.1
940	USH00384690	Tmax_adj	1973	2025	7	34.5367	-80.5972	138.1	32.3	31.7	30.6
925	USH00380074	Tmax_adj	1973	2025	7	33.4925	-81.6958	150	33.3	31.5	30.5
937	USH00383468	Tmax_adj	1973	2025	7	33.3619	-79.2239	3	31	30.4	29.5
939	USH00383754	Tmax_adj	1973	2025	7	34.1997	-82.1711	187.5	33.5	32.6	31.9
938	USH00383747	Tmax_adj	1973	2025	7	34.8847	-82.2239	287.4	32.1	32.1	30.9
946	USH00387631	Tmax_adj	1973	2025	7	33.9978	-81.7747	146.3	32.8	32.9	31.8
941	USH00384753	Tmax_adj	1973	2025	7	33.6775	-79.8283	22.9	34.3	33.6	31.5
945	USH00386527	Tmax_adj	1973	2025	7	33.4886	-80.8733	54.9	31.8	31.3	30.1
949	USH00388440	Tmax_adj	1973	2025	7	33.9364	-80.3569	53.9	32.7	32.4	31.2
952	USH00389350	Tmax_adj	1973	2025	7	34.9381	-81.0317	210.3	32.4	32.3	30.3
975	USH00396947	Tmax_adj	1973	2025	7	44.115	-103.2828	1052	28.7	31.4	30
972	USH00395891	Tmax_adj	1973	2025	7	43.8894	-100.7078	707.1	31.6	34.8	32.5
959	USH00391972	Tmax_adj	1973	2025	7	43.9611	-101.8606	735.8	32.3	35.3	33
956	USH00390128	Tmax_adj	1973	2025	7	43.6567	-97.7853	412.4	28.8	32.2	31.6
966	USH00394007	Tmax_adj	1973	2025	7	43.4378	-103.4739	1085	30.5	34	33.2
955	USH00390043	Tmax_adj	1973	2025	7	43.4892	-99.0631	512.1	30.6	34.8	33.7
957	USH00391392	Tmax_adj	1973	2025	7	43.3056	-96.5917	410	28	32.3	31.1
968	USH00394516	Tmax_adj	1973	2025	7	43.9072	-99.8628	518.2	32.9	35.5	34.9
960	USH00392429	Tmax_adj	1973	2025	7	45.0481	-101.5992	723.9	33.8	36.7	35.3
974	USH00396597	Tmax_adj	1973	2025	7	44.3814	-100.2856	531	33	35.9	34.8
963	USH00393029	Tmax_adj	1973	2025	7	44.0422	-98.07	374.9	29.4	32.7	33.2
961	USH00392797	Tmax_adj	1973	2025	7	45.7644	-99.6353	566.9	29.1	31.3	29.9
954	USH00390020	Tmax_adj	1973	2025	7	45.4433	-98.4131	395.3	30.8	32.9	33.5
964	USH00393217	Tmax_adj	1973	2025	7	44.0569	-99.0719	524.3	30.9	34.5	34.1
970	USH00395481	Tmax_adj	1973	2025	7	43.2358	-97.5714	403.6	30.6	34	33.4
973	USH00396170	Tmax_adj	1973	2025	7	44.4419	-100.4175	506	33.2	37	35.7
971	USH00395536	Tmax_adj	1973	2025	7	45.2836	-96.6689	349	30.2	32.8	33.2
977	USH00398932	Tmax_adj	1973	2025	7	44.9047	-97.1494	532.8	28.8	32.1	32
958	USH00391739	Tmax_adj	1973	2025	7	44.8817	-97.7325	549.9	29.1	32.3	31.6
978	USH00399442	Tmax_adj	1973	2025	7	43.4978	-100.4794	664.5	32.5	36.3	34.7
962	USH00392927	Tmax_adj	1973	2025	7	45.0364	-99.1342	478.5	31.7	34.5	33.1
976	USH00398622	Tmax_adj	1973	2025	7	42.7625	-96.9194	362.7	29.5	34.5	33
969	USH00395456	Tmax_adj	1973	2025	7	45.155	-98.5825	396.8	31	33.9	33.3
965	USH00393832	Tmax_adj	1973	2025	7	44.5222	-99.455	576.1	31.8	34.8	34.4
967	USH00394037	Tmax_adj	1973	2025	7	44.0122	-97.5242	474.9	30.6	34.1	33.2
982	USH00402202	Tmax_adj	1973	2025	7	36.0147	-85.1311	551.7	26.5	26.7	26.6
986	USH00405187	Tmax_adj	1973	2025	7	35.4139	-86.8086	239.9	31.6	32.2	31.2
985	USH00404561	Tmax_adj	1973	2025	7	35.6214	-88.8456	121.9	32.7	32.7	31.6
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979	USH00401790	Tmax_adj	1973	2025	7	36.5467	-87.3567	116.4	33.4	33.2	33.2
987	USH00405882	Tmax_adj	1973	2025	7	35.6717	-85.7811	286.5	30.4	30.9	30
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990	USH00407884	Tmax_adj	1973	2025	7	36.4161	-82.9839	413	29.9	30.2	30.4

988	USH00406371	Tmax_adj	1973	2025	7	35.9203	-86.3728	163.1	31.2	32.5	31.2
980	USH00402024	Tmax_adj	1973	2025	7	34.9939	-84.3758	442	31	30.2	30.2
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981	USH00402108	Tmax_adj	1973	2025	7	35.5497	-89.7	117.3	32.1	32.3	31.1
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997	USH00410493	Tmax_adj	1973	2025	7	31.7414	-99.9764	534.9	32.3	36	32.5
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1017	USH00413280	Tmax_adj	1973	2025	7	30.9072	-102.9153	926	34.9	36.6	33.7
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1039	USH00418433	Tmax_adj	1973	2025	7	32.71	-100.9111	706.8	33.8	35.7	31.8
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1033	USH00416892	Tmax_adj	1973	2025	7	31.4167	-103.5	795.5	34.4	36.8	33.1
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1004	USH00411138	Tmax_adj	1973	2025	7	31.7383	-98.9456	426.7	34.7	35.2	33.9
1012	USH00412679	Tmax_adj	1973	2025	7	28.7567	-100.4789	247.5	34.2	36.7	32.5
1011	USH00412598	Tmax_adj	1973	2025	7	32.0628	-98.3047	447.4	32.6	35.8	31.9
1014	USH00412906	Tmax_adj	1973	2025	7	28.0239	-99.3508	176.8	35.2	36.4	33.9
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1015	USH00413063	Tmax_adj	1973	2025	7	27.1394	-98.12	42.4	36.8	34.7	34
1020	USH00413873	Tmax_adj	1973	2025	7	29.4706	-96.9397	83.8	33.5	35.1	33.5
1030	USH00416135	Tmax_adj	1973	2025	7	34.2192	-102.7328	1167	30.8	32.5	28.3
1038	USH00418201	Tmax_adj	1973	2025	7	32.7136	-102.6597	1017	32.9	35	29.9
1029	USH00415875	Tmax_adj	1973	2025	7	35.7	-100.6436	839.7	32.3	34.5	30.1
998	USH00410498	Tmax_adj	1973	2025	7	30.9844	-103.7403	981.5	32.8	35.3	30.9
1022	USH00415018	Tmax_adj	1973	2025	7	31.0717	-98.1847	314.6	33.3	35.6	33.3
1010	USH00412266	Tmax_adj	1973	2025	7	29.0567	-96.2319	21.3	33.5	33.8	33.1
1007	USH00412015	Tmax_adj	1973	2025	7	27.7742	-97.5122	13.4	33.8	33.8	33.4
1040	USH00418692	Tmax_adj	1973	2025	7	36.3372	-102.0753	1126	32.8	34.2	31.2
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1027	USH00415707	Tmax_adj	1973	2025	7	31.1333	-102.2217	757.7	34.5	36.7	33.9
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995	USH00410144	Tmax_adj	1973	2025	7	27.7283	-98.0678	61.3	35.2	36	34.6
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1002	USH00411000	Tmax_adj	1973	2025	7	35.5303	-102.2564	972.6	33.2	34.1	31.1
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1024	USH00415272	Tmax_adj	1973	2025	7	30.7425	-98.6542	310.9	34.4	36.3	33.7
1001	USH00410902	Tmax_adj	1973	2025	7	29.7978	-98.7347	440.1	32.6	35.2	33
1008	USH00412019	Tmax_adj	1973	2025	7	32.1078	-96.4747	125.9	33.7	36.5	34
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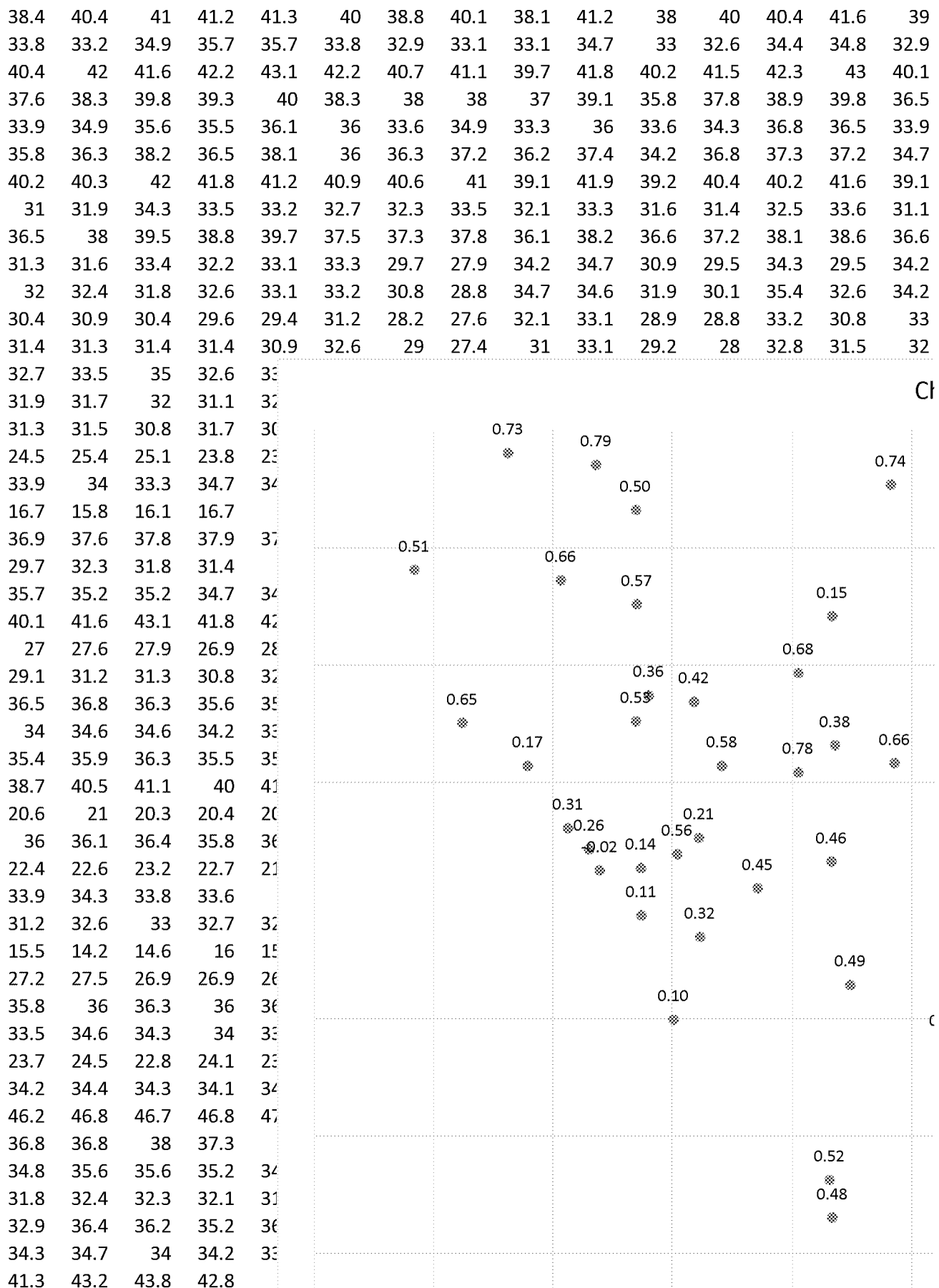
1028	USH00415869	Tmax_adj	1973	2025	7	31.7003	-96.5142	161.5	33.1	35.9	33.9
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1062	USH00425733	Tmax_adj	1973	2025	7	38.5744	-109.5458	1243	34	34.5	34.2
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1053	USH00423418	Tmax_adj	1973	2025	7	38.9906	-110.1544	1241	34.9	35.6	35.5
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1073	USH00427729	Tmax_adj	1973	2025	7	39.6847	-111.2047	2655	22	22.9	22.5
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1067	USH00426601	Tmax_adj	1973	2025	7	37.8239	-112.4419	2021	27.9	27	27.3
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1065	USH00426135	Tmax_adj	1973	2025	7	39.7122	-111.8319	1563	31.7	32.8	32.7
1069	USH00427260	Tmax_adj	1973	2025	7	38.7619	-112.0775	1615	31.3	31.6	30.6
1048	USH00422253	Tmax_adj	1973	2025	7	40.1678	-110.395	1683	29.5	31.2	30.2

1051	USH00422828	Tmax_adj	1973	2025	7	38.9664	-112.3278	1561	31.5	32.7	32.4
1061	USH00425477	Tmax_adj	1973	2025	7	38.45	-112.2292	1801	32.1	31	30.7
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1092	USH00440993	Tmax_adj	1973	2025	7	37.7092	-78.2886	68.6	32.3	31.9	30.2
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1091	USH00440766	Tmax_adj	1973	2025	7	37.2017	-80.4133	640.1	26.9	26.6	26.5
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1108	USH00449151	Tmax_adj	1973	2025	7	37.3017	-76.7039	21.3	30.1	30.1	29
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1096	USH00442245	Tmax_adj	1973	2025	7	36.5869	-79.3886	125	31.8	31.8	30.7
1107	USH00448062	Tmax_adj	1973	2025	7	38.1814	-79.0903	499.9	29.5	29.4	28.3
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1104	USH00446626	Tmax_adj	1973	2025	7	36.7381	-82.9967	448.1	30.2	30.8	31.2
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1084	USH00431243	Tmax_adj	1973	2025	7	43.3847	-72.5989	256.6	28.6	27.3	28.7
1086	USH00431580	Tmax_adj	1973	2025	7	43.9572	-73.2106	105.2	27.5	26.4	28.7
1085	USH00431360	Tmax_adj	1973	2025	7	43.9833	-72.45	243.8	27.1	25.9	27.9
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1083	USH00431081	Tmax_adj	1973	2025	7	44.4681	-73.1503	100.6	28.3	27.5	30.6
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1148	USH00458773	Tmax_adj	1973	2025	7	45.6778	-122.6519	64	24.8	22.8	24
1132	USH00455946	Tmax_adj	1973	2025	7	48.9114	-117.8069	423.7	31.4	27.5	30
1149	USH00458928	Tmax_adj	1973	2025	7	46.1	-118.2833	355.4	31.4	29.3	31.6
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1133	USH00456039	Tmax_adj	1973	2025	7	47.3286	-118.6997	466.3	30.8	28.6	31.3
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1138	USH00456789	Tmax_adj	1973	2025	7	46.7564	-117.1911	775.7	30.1	27.5	29.3
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1119	USH00451666	Tmax_adj	1973	2025	7	48.5467	-119.7492	707.1	28.8	25.4	29.2
1144	USH00457773	Tmax_adj	1973	2025	7	47.5414	-121.8361	134.1	24.1	21.9	23.3
1111	USH00450587	Tmax_adj	1973	2025	7	48.7178	-122.5114	4.6	20.4	19.1	20.7

1121	USH00452007	Tmax_adj	1973	2025	7	47.6503	-118.1456	743.7	29.1	26.3	29.6
1153	USH00459376	Tmax_adj	1973	2025	7	48.4539	-120.1936	534.9	30.5	26.6	30.4
1152	USH00459238	Tmax_adj	1973	2025	7	47.7569	-118.6775	679.7	29.6	27.5	30.5
1122	USH00452030	Tmax_adj	1973	2025	7	46.3153	-118.0022	474.6	31.2	29.3	31.2
1117	USH00451504	Tmax_adj	1973	2025	7	47.1856	-120.9469	585.2	27.7	25.5	28
1123	USH00452505	Tmax_adj	1973	2025	7	46.9692	-120.54	451.1	29.9	28	30.2
1142	USH00457458	Tmax_adj	1973	2025	7	47.65	-122.3	5.8	24.9	23.4	24.5
1143	USH00457507	Tmax_adj	1973	2025	7	48.4958	-122.2355	18.3	23.1	21.5	23.4
1120	USH00451939	Tmax_adj	1973	2025	7	47.3706	-123.16	6.4	23.7	21.6	23.2
1114	USH00451233	Tmax_adj	1973	2025	7	47.4144	-121.7561	475.5	20.5	18.8	19.8
1113	USH00450945	Tmax_adj	1973	2025	7	47.1694	-122.0036	208.8	24.5	22.6	24
1115	USH00451276	Tmax_adj	1973	2025	7	46.72	-122.9528	56.4	25.2	22.9	24
1116	USH00451484	Tmax_adj	1973	2025	7	48.9672	-122.3292	19.5	24	22.1	23.7
1134	USH00456096	Tmax_adj	1973	2025	7	48.6117	-122.8064	24.4	20.4	19.1	20
1150	USH00459012	Tmax_adj	1973	2025	7	47.6478	-120.0697	798.6	28.6	26.4	29.6
1140	USH00457059	Tmax_adj	1973	2025	7	47.1139	-118.3772	557.8	31.4	28.5	31.7
1151	USH00459074	Tmax_adj	1973	2025	7	47.4247	-120.315	195.1	32.4	29.6	32.6
1135	USH00456610	Tmax_adj	1973	2025	7	46.4686	-117.5894	579.1	30.6	27.9	31.5
1125	USH00452914	Tmax_adj	1973	2025	7	47.9558	-124.3539	106.7	20.3	19.8	22.1
1136	USH00456624	Tmax_adj	1973	2025	7	48.1139	-123.4317	27.4	18.6	17.5	18.9
1112	USH00450729	Tmax_adj	1973	2025	7	48.9775	-122.7928	18.3	21.2	20.1	22
1131	USH00455224	Tmax_adj	1973	2025	7	47.1358	-122.2558	176.5	24.6	22.5	24.3
1124	USH00452675	Tmax_adj	1973	2025	7	47.9753	-122.195	18.3	24.6	23.2	24.3
1128	USH00454748	Tmax_adj	1973	2025	7	46.3675	-124.0378	7.6	17.3	17.4	17.9
1139	USH00456914	Tmax_adj	1973	2025	7	46.6533	-123.73	9.1	21.7	21.6	21.6
1110	USH00450008	Tmax_adj	1973	2025	7	46.9658	-123.8292	3	20	19.8	20
1168	USH00470991	Tmax_adj	1973	2025	7	44.8558	-88.9889	329.2	27.2	28.2	27.7
1167	USH00470349	Tmax_adj	1973	2025	7	46.5831	-90.9678	198.1	25.9	27.8	27.7
1172	USH00473405	Tmax_adj	1973	2025	7	44.1192	-89.5342	331.9	28.1	28	27.7
1178	USH00475516	Tmax_adj	1973	2025	7	45.8864	-89.7322	488	24.2	25.4	24.9
1181	USH00476208	Tmax_adj	1973	2025	7	44.8836	-87.9539	201.2	27.3	28.2	27.7
1169	USH00471078	Tmax_adj	1973	2025	7	42.6181	-89.3861	240.8	28.9	29.9	28.2
1177	USH00475474	Tmax_adj	1973	2025	7	43.0719	-88.0294	221.3	27.8	28.3	28
1185	USH00476922	Tmax_adj	1973	2025	7	42.7022	-87.7861	181.4	26.7	27	27
1176	USH00475255	Tmax_adj	1973	2025	7	45.1308	-90.3439	448.1	26.5	27.5	27.5
1170	USH00472001	Tmax_adj	1973	2025	7	42.6708	-90.1183	292.6	28.2	29.2	28.4
1182	USH00476330	Tmax_adj	1973	2025	7	44.0206	-88.5581	228.6	28.3	28.7	28.5
1183	USH00476718	Tmax_adj	1973	2025	7	43.5278	-89.4342	236.2	28.2	29.1	27.9
1186	USH00478027	Tmax_adj	1973	2025	7	45.8236	-91.8761	335.3	27	29	29.2
1175	USH00475120	Tmax_adj	1973	2025	7	44.6319	-90.1314	377	26.3	28.2	27
1171	USH00472839	Tmax_adj	1973	2025	7	43.7961	-88.4506	231.6	28	28	28.3
1187	USH00478110	Tmax_adj	1973	2025	7	44.9686	-90.9389	332.2	27.5	29.3	27.7
1180	USH00475932	Tmax_adj	1973	2025	7	44.3589	-88.7189	243.8	28.9	29.7	28.6
1179	USH00475808	Tmax_adj	1973	2025	7	44.5297	-90.6383	310.9	28.6	29.2	27.8
1173	USH00474546	Tmax_adj	1973	2025	7	42.8278	-90.7889	317	27.6	29.3	27.8
1189	USH00478919	Tmax_adj	1973	2025	7	43.1742	-88.7364	251.5	28.8	29.1	28.2
1174	USH00475017	Tmax_adj	1973	2025	7	44.0869	-87.6522	178	25.7	26.1	26.5

1188	USH00478827	Tmax_adj	1973	2025	7	43.5594	-90.8761	382.5	27.6	29	28.5
1184	USH00476827	Tmax_adj	1973	2025	7	43.0514	-91.135	200.6	29.3	30.8	30.2
1159	USH00465707	Tmax_adj	1973	2025	7	39.4019	-77.9844	162.8	28.8	29.1	27.6
1157	USH00465224	Tmax_adj	1973	2025	7	37.8564	-80.4042	701	27.1	26.7	26.7
1165	USH00469610	Tmax_adj	1973	2025	7	37.6731	-82.2761	231.6	31	30.8	30.7
1158	USH00465626	Tmax_adj	1973	2025	7	39.5436	-80.4647	335.3	28.3	28.3	28.9
1160	USH00466867	Tmax_adj	1973	2025	7	39.1031	-79.6658	556.6	28.5	27.8	27.9
1161	USH00466989	Tmax_adj	1973	2025	7	38.6667	-80.2	877.8	24.9	24.5	24.4
1156	USH00463544	Tmax_adj	1973	2025	7	38.9339	-80.8325	216.4	29.3	28.8	29.5
1154	USH00461220	Tmax_adj	1973	2025	7	38.98	-80.2206	443.5	29.9	28.8	29
1155	USH00461330	Tmax_adj	1973	2025	7	39.2047	-81.1572	231.6	29.6	28.9	29.4
1164	USH00469368	Tmax_adj	1973	2025	7	40.2753	-80.6119	201.2	29	29.1	28.9
1162	USH00467029	Tmax_adj	1973	2025	7	37.5744	-81.5356	390.1	27.9	27.2	27.2
1163	USH00468384	Tmax_adj	1973	2025	7	38.8008	-81.3619	287.4	28.3	28	28
1166	USH00469683	Tmax_adj	1973	2025	7	38.5278	-81.9153	186.2	31	30.5	30.5
1195	USH00481840	Tmax_adj	1973	2025	7	44.5219	-109.0633	1549	24.7	24.8	25.7
1212	USH00487845	Tmax_adj	1973	2025	7	41.5942	-109.0653	2055	24.5	27.7	27.9
1201	USH00485345	Tmax_adj	1973	2025	7	44.5619	-110.3986	2399	20.1	22.4	21.5
1217	USH00489770	Tmax_adj	1973	2025	7	44.0108	-107.9686	1238	30.2	33.3	32.9
1218	USH00489905	Tmax_adj	1973	2025	7	44.9767	-110.6964	1899	26.2	27.9	27.1
1200	USH00484065	Tmax_adj	1973	2025	7	41.5314	-109.4767	1852	28.5	30.3	29.5
1205	USH00486440	Tmax_adj	1973	2025	7	43.8567	-110.5889	2072	21.4	24.1	23.2
1209	USH00487260	Tmax_adj	1973	2025	7	42.8797	-109.8642	2193	23.6	25.4	25.5
1199	USH00483100	Tmax_adj	1973	2025	7	41.265	-110.9508	2080	24	25.7	25.4
1190	USH00480140	Tmax_adj	1973	2025	7	43.7728	-111.0339	1962	24.2	26.7	26.3
1193	USH00481675	Tmax_adj	1973	2025	7	41.15	-104.8167	1868	25.8	28.7	27.2
1214	USH00488160	Tmax_adj	1973	2025	7	44.8406	-106.8383	1143	28.8	32.4	29.9
1210	USH00487388	Tmax_adj	1973	2025	7	44.7764	-108.7592	1332	28	30.6	29.9
1198	USH00482715	Tmax_adj	1973	2025	7	43.5397	-109.6553	2120	23.1	26.7	26.2
1202	USH00485415	Tmax_adj	1973	2025	7	41.3125	-105.6744	2215	25.4	27.6	27.6
1191	USH00480540	Tmax_adj	1973	2025	7	44.3789	-108.0314	1170	32	34.4	34.4
1213	USH00487990	Tmax_adj	1973	2025	7	41.4528	-106.8053	2070	26.7	28.1	28.2
1197	USH00482595	Tmax_adj	1973	2025	7	43.2281	-108.9489	1699	28.9	31.8	30.4
1215	USH00488995	Tmax_adj	1973	2025	7	42.0803	-104.2236	1249	29.6	33.3	32.5
1194	USH00481730	Tmax_adj	1973	2025	7	41.7592	-104.8219	1617	28.8	31.2	30.6
1196	USH00481905	Tmax_adj	1973	2025	7	44.8711	-104.1533	1061	28.2	31.3	31.1
1203	USH00485830	Tmax_adj	1973	2025	7	42.7506	-104.4811	1551	27.6	30.5	29.5
1211	USH00487760	Tmax_adj	1973	2025	7	43.0308	-108.3742	1510	29.1	32.4	31.7
1204	USH00486195	Tmax_adj	1973	2025	7	43.4131	-106.2772	1481	29.2	31.6	31.4
1208	USH00487240	Tmax_adj	1973	2025	7	41.1722	-104.1583	1579	28.4	31.9	31.1
1207	USH00487115	Tmax_adj	1973	2025	7	43.2458	-108.6942	1658	27.7	30.6	29.4
1216	USH00489615	Tmax_adj	1973	2025	7	42.1106	-104.9492	1414	28	31.5	30.9
1192	USH00480552	Tmax_adj	1973	2025	7	42.6339	-106.3775	1832	25.9	28.7	29
1206	USH00486660	Tmax_adj	1973	2025	7	43.8581	-104.2136	1315	30.2	32.9	32.2

1976	1977	1978	1979	1980	1981	1982	1983	1984	1985	1986	1987	1988	1989	1990
31.5	34.1	34.1	30.7	36.6	33	32.8	33.8	32.1	32.7	35.6	33.6	32.8	31.2	33
31.8	34.5	33	31.5	36.2	35.6	31.9	33.5	31.6	31	34.7	32.7	31.9	30.5	32
31.7	33.7	33.7	31.1	34.1	33.5	32.2	33.6	31.5	31.5	34.1	33.3	31.2	31.3	32.6
34.2	33.5	33.5	32.5	34.1	34.2	32.4	34	32.3	31.8	35	33.1	32	30.6	31.8
32.4	33.6	31.8	31.5	33.8	33.5	32	34	31.9	31	33.9	33.1	31.8	31.4	32.1
31.9	34.9	32.8	31.8	34.9	34	32.1	34.4	32.5	31.8	35.4	33.8	32.5	31.3	33.3
33.3	34.4	34.4	32.3	35.6	34.6	33.1	33.7	32.4	32	35.3	33.7	31.8	31.7	32.3
33.4	35	35.4	32.9	35.3	35	34.4	35	33.5	33.7	36	35	33	32.7	33.8
30.6	33.4	33	29.3	35.1	33	31.6	32.5	30.7	30.8	34	33	32.1	30.2	32
33.2	34.1	33	31.8	34.2	34	32.8	32.8	31.9	31.2	33.8	32.9	31.4	30.8	32.7
33.5	34.2	34.1	31.9	36.1	35	32.8	33.4	32.3	31.6	35.4	33.5	32	31.2	32.9
32.1	33.8	32.8	31.3	34.5	33.1	32.2	33.2	31.3	31.2	34.6	33.4	30.6	30.8	32.2
28.5	31.5	31.1	27.9	33.2	30.8	29.8	31.1	28.8	29	32.6	31.1	29.8	28.4	30.4
31.1	34.5	33.1	30	35	32.8	32	32.5	31.3	30.5	33.5	32.1	31	30.1	31.5
33.3	35	34.9	32.5	35.6	35.3	33.5	34.9	33.2	33.5	35.5	34.3	32.6	32.1	34.1
30.9	33	35.1	30.6	37.3	31.7	31.7	32.9	33.1	33.1	34.5	31.6	32.6	29.8	31.9
32.5	33.5	34.9	31.5	38.7	33.7	33.2	33	32.7	34.2	36.2	32.6	32.2	27.6	32.9
30.2	32	33	29.5	35.9	31	30.6	32	32.3	32.9	33.4	30.7	32	29.3	30.9
32	33.1	35.2	30	36.3	31.1	31.2	31.1	31.1	31.6	33.7	31	31.4	29.7	31.6
32.8	33	35.8	31.1	38	33.9	32.5	32.9	33.1	34.7	35.2	32	32.4	30.3	33.1
33.2	33	35.8	31.6	37.9	33.6	33.1	33	33.7	35.1	35.8	32.8	32.3	30.1	33
32.4	34.5	37.2	32	37.4	33.7	33.8	33.3	32.5	33.3	34.8	33.3	33.4	31.2	33
32.6	33.9	35.8	32.5	37.6	33.5	33.5	34.1	32.8	33.3	35.7	33.2	33.2	31	33.3
31.4	32.4	33.5	30.9	37.9	32.5	31.9	33.2	32.9	33	35.3	31.6	30.8	29.6	32.1
32.4	33.6	34.2	31.4	36.4	33.6	32.9	33.2	31.7	32.8	35.3	32.2	31.9	30.6	32.9
31.1	33.3	34.7	30.2	38	32.5	32.1	33.6	32.1	33.6	34.4	32.5	31.5	29	31.4
32.6	33.4	33.5	31.5	37.9	33.6	33.8	35.6	33.3	35	35.7	32.7	32.8	31.4	33.7
32.2	32.7	33.3	31.3	37.2	33	33.1	34.1	32.2	33.8	35.2	32	31	29.7	32.9
32.3	33.5	33.9	31.6	38.6	33	33.2	35	32.6	33.7	35	32	32.1	31.2	33.6
31.8	33.3	33.9	32	35	32.7	33.1	34	31.4	32.5	34.8	32.5	32.2	30.5	33.5
30.1	30.3	32	31.7	31.6	30.6	30.4	29.8	30.9	31.3	30.9	31.5	32.5	33.4	31.7
30.5	29.3	31.4	31.6	31.8	30.4	30.3	30.7	28.9	31.5	29	31	31.2	32.2	30.8
35.9	38.2	37.8	37.7	38.4	36.7	36	36.9	35.9	37.1	35.2	36.9	37.7	38.5	36.6
30	31.6	33.7	33	34.5	31.2	31.8	31.9	31.3	33.1	29.8	32.8	31.6	33.4	30.2
38.3	39	39.7	38.9	39.8	38.8	37	37.8	36.4	38.6	36.5	37.4	38.6	39.3	37.1
26.8	27	28.7	28.9	27.8	26.8	25.9	26.2	25.7	28.1	26.8	26.6	29.2	29.8	27.3
31.1	31.4	32.7	33.2	33.2	32.3	31.2	31.2	30	32.7	29.9	31.4	34	34.3	32.2
34.7	36.9	37.6	37.6	37.9	35.2	35.1	37.1	34.6	37.2	35.2	37.3	37.1	38.2	34.3
25.6	26.4	27.6	28.4	26.8	23.8	25.6	26.8	24.6	26.4	24.6	25.1	27.2	27.8	25.5
37.5	38.8	39.5	39.9	40.4	39.3	38.5	39.1	37.5	40.1	37.8	39.1	40	40.8	37.8
37.2	36.4	38.2	38.3	38.7	37.5	36.8	36.2	37.1	37.7	35.9	35.8	38.6	39.2	37.5
30.2	30.9	32.2	32.3	32.8	31.4	30	30.8	29.6	31.7	29.1	30.6	32.1	33.1	30.3
26.1	26.8	28.4	29.2	28.3	27.6	27	27.3	26.1	27.8	26	26.7	28.5	28.8	26.6
39.6	41.9	42.5	41.8	41.9	41.5	40.6	41.5	39.5	41.6	39.4	40.1	40.7	42.4	41.2
40.8	42.4	42.9	42.5	43.2	42.5	40.8	41.6	40.5	42.4	40.3	40.9	42.6	43.7	41.3
30.7	32.2	35.4	34.5	34.8	31.7	32.4	32.7	31.4	33.8	31.3	33.7	33	34.4	30.8



41.4	42.8	43.4	42.5	43.5										
32.4	33.3	31.5	32.2											
26.8	28.3	26.7	26.2	26.5										
34.1	33.8	34	33.2											
31.6	33.1	32.8	32.3	31.5										
30.3	31.1	30.8	31.1	29.5										
22.8	22.7	22.6	22.6	23.5										
27.1	28.2	27.6	25.7	26.5										
29.5	30.7	30.9	30.6	31.5										
34.4	34.5	35	34.5	34.5										
32.8	34.5	34.1	33.1	33.5										
31.9	32	30.7	31.6	30.5										
27	27.4	27.3	28		-125	-124	-123	-122	-121	-120				
22.2	22.4	21.5	23	22.2	22.3	21.7	23.9	23.7	23.1	22.1	23.2	24.1	24	22.1
23.7	23.7	22.4	23.8	23.3	22.9	22.6	25.2	23.8	24.8	21.5	23.6	22.8	23.6	23.5
26.2	27	27.1	27.5	27.2	27	26.9	27.3	29.6	28.2	27.4	24.6	27.9	28.1	26.7
24.1	23.6	24.3	24	27.5	25.7	26.3	25.8	25.8	25.2	24.5	26.5	26.8	26.8	24.4
32	30.6	31.6	30.9	31.2	30.4	29.8	29.1	29.9	29.9	28.5	29.4	31.4	31.9	30.2
32.3	31.1	32.3	31.9	32.6	31.3	31	30	31	31.5	29.1	30.6	32.9	32.7	30.7
25.7	24.5	25.5	25.5	26.2	25	25	25.3	24.9	24.8	24	25.7	25.7	27	25.5
35.5	33.5	35.5	34.8	35.3	34.8	34.9	33.1	33.9	34.2	32.4	32.7	36.2	36.2	34.2
23.1	23.4	24.2	23.8	23.9	22.3	21.9	22.7	22.5	22.3	20.9	22.7	23.8	23.9	22
26.9	26.2	27.3	26.4	28.1	26.3	26.5	26.1	26	24.8	24.1	26.5	26.4	27.3	24.9
30.2	29.9	30.5	30	31	29.8	29.5	29.8	29.9	29.2	29.3	30.3	29.9	30.7	27.4
24.9	22.9	25.7	25.1	26.1	23.3	24.1	23.7	23.6	23.4	21.5	23.6	23.7	25.2	22.5
30.3	30.4	31.6	30	32.5	29.9	30.1	30.3	29.9	29.4	28.9	31.1	29.5	30	27.7
31.4	31.7	32.7	30	32.7	29.4	30.8	32.3	31.2	31	31.6	31.3	31.5	31.2	30.2
30.7	30.5	30.5	30.2	31.3	29.5	28.7	29.5	30.4	28.5	28.6	29.9	29.8	30.6	27.1
30.6	30.8	31.8	29.5	32.4	30.5	30.8	32.7	31	30.2	30.8	31.9	29.9	29.4	28.7
25.9	26.1	27.5	27.6	27.9	26.1	25.6	26.7	25.9	25.7	24.8	27	25.3	27.5	24.3
29	28.5	28.3	28.4	29.2	27.4	27.1	26.5	27.1	27	25.7	26.7	28.9	29.7	26.3
28.9	28.9	27.3	29.1	28.5	28.3	28.7	28.7	28.8	27.4	27.7	28.2	28.9	30.2	26.7
30.9	31.9	32.8	30.6	33.5	31.4	30.7	32.6	30.8	30.4	30.7	32	31.4	30.9	28.6
33.7	34.6	35	33.3	36.2	33.7	33.5	35.2	33.4	33.4	34.2	34.5	33.9	33.4	32.3
32.6	32.8	33.5	31.7	35.4	33	32	34.2	32.4	33.8	33.4	32.5	32.9	31.9	31.2
33.5	33.4	33.8	32.2	34.6	31.6	32.4	32.9	32.9	32.8	33.5	32.3	33.2	33.9	30.1
26.4	25	25.7	25.5	27.3	23.3	25.7	25.9	24.6	24.6	23.5	25.5	25.4	27.1	24.2
31.3	30.7	32.6	32.4	33.8	31.6	31.4	31.9	31.5	30.9	31.4	31.8	31.5	32.1	29.6
34.8	35.2	36.4	34.2	38.2	34.7	32.8	36.5	35.7	34.4	35.5	35.8	35.1	34.3	32
34.2	35.5	35.9	34.1	38.3	34.1	33.7	35.4	34.4	33.8	34.4	34	34	33.7	32.5
26	28.1	26.5	28.1	28.9	27.8	28.2	28.9	26.7	27.4	26.7	28.5	29.1	27.3	28
25.8	28.6	27.7	28.1	29.2	27.7	28.4	29.6	26.8	27.7	26.9	28.8	29	27.2	26.2
25.1	26.1	25.5	26.8	27.1	27.1	26.9	27.7	25	26.5	25.1	26.5	26.6	25.7	25.9
26.5	27.6	25.7	27.5	27.8	27.9	28.3	28.6	26	26.5	25.6	27.1	26.8	26.2	26.2
26.4	28.5	26	26.8	29	27.9	28.5	30.2	27.5	28.2	28.7	29.9	30.7	27.2	27.8
28.7	31	29	29.1	31.9	29.9	30.4	32.3	29.2	30.1	30.7	31.6	32.3	29.5	30
27.9	30.2	27.9	27.7	30.3	29.1	29	30.8	28.2	28.9	29.4	32.1	32	28.8	30.3

30.3	32.2	29.3	28.5	31.5	30.5	31.2	32.3	28.9	30.2	31.6	32.3	32.3	29.4	30.8
29.4	31	28.6	28.1	30.9	29.9	29.6	32	27.9	30	31	31.6	31.5	28.7	30.6
31.7	32.1	32.2	32.6	32.1	32.3	32.3	32.8	32	31.6	31.8	32.9	32.6	32.5	32.7
33.5	34.5	32.1	32.5	33.8	33	31.9	33.7	30.9	32.1	34.1	34	32.9	32.9	34.1
31	30.6	31.1	31.5	31.7	33.1	31.1	31.8	30.1	30.6	31.4	31.6	30.8	32.3	31.1
32.8	33	33.2	33.7	33.8	34.2	33.2	33.4	32.5	32.5	33.3	33.6	33	33.1	33.1
30.9	30.3	31.2	31.2	31.5	31.8	31.9	30.6	29.8	30.9	31.8	31.9	31.5	31.3	32.2
31.1	31.7	30.5	32.4	32.1	32.7	31.4	32.1	30.8	31.2	31	32.7	31.7	32	31.7
31.2	31.6	30.3	31.4	31.3	31.9	30.5	31.5	29.2	31	32.5	32.6	31	31.2	32
32	31.2	31.5	32.4	32.8	32.6	32	32.1	31.4	31.6	32	32.3	31.7	32	32.8
31.3	31.7	30.9	32.4	32.4	33	31.9	32.5	31.1	31.6	32.8	32.1	30.7	32.6	31
33.3	33.5	32.5	32.2	35	34.4	32	35.1	31.7	32	35.4	34.2	32.3	32.1	34.3
32.7	32.2	32.6	34.1	33.1	33.7	33	33.1	32.1	32.1	32.9	33.7	33.4	33	33.6
32.6	33.7	32.2	32.9	32.6	34.1	32.3	33.7	31.4	32.3	34	33	31.9	30.7	32.8
32.5	32.5	31.9	32.9	32.7	33	32	33	31.5	31.9	32.3	32.7	32.2	33	32.5
32	31.9	32	32.5	32.4	32.8	32.3	33	31	31	32	31.8	31.6	31.4	32.3
31.6	32.5	30.7	32.3	33.1	33.1	32.1	32	31	31.3	33	32.2	31.7	32.2	32.9
32	33.1	32.5	33.8	33.9	34.6	33.4	33.5	32.3	32.5	32.7	33.6	32.7	33.1	32.6
33.8	33.8	32.7	33.4	33.9	34.6	33	33.6	31.9	32.2	34.8	33.2	32	32.8	33.4
33.8	34.5	32.7	33.5	33.7	34	31.5	33.9	31.6	32.3	34.9	34.1	32.8	31.9	34.3
31.6	32.1	32	31.7	34.1	33.5	31.6	33.1	30.7	31.2	34.1	33.5	31.4	31.4	32.3
32.7	33.3	33.3	34	34.3	35.7	33.5	34.4	33	33.5	33.2	33.3	33	33.1	34.4
33.9	34.5	33.2	34.6	34.2	35	33.1	33.9	32.9	33.5	32.6	34.2	32.7	33.6	33.6
33.9	34	32.7	34.3	34.5	34.9	33.1	33.8	32.8	33.6	33.6	34.4	33.3	33.2	33.1
28.5	31.1	30.4	27	31.6	31	29.6	31.7	28.1	28.2	34.3	31.9	30.6	29.1	30.6
32.1	34	34.6	31.7	34.7	34.1	31.9	34.6	31.1	31.3	37.5	33.8	32.6	30.5	34.9
32.8	34	32.7	32.3	34.3	34.2	31.7	34	31.7	33.2	35.9	34.2	33.4	33.3	34.3
30	35	33	29.8	34.2	33.2	31.3	32.8	30.1	30	35.2	33.3	32.6	29.8	32.3
32.3	33.8	32.3	32.1	33.7	34.9	31.7	33.5	30.6	32.2	34.8	33.8	32.3	31.3	33.6
34.5	35.2	33	31.4	35.2	35.9	32.5	34.6	31.3	32.1	36	33.9	32.5	31.8	34
33.4	33.8	33.2	33.3	33.7	35.3	32.7	33.2	32	32.9	34.4	33.1	32.3	32.5	33.5
32.7	34.9	32.6	31.5	35.8	34.1	32.1	34.2	31	32	35.7	33.8	34.2	31	34.3
34.5	35.8	33.9	33.6	35.8	36.5	33.6	35	33.6	33.3	36.9	34.8	34.6	33.6	36.1
34.1	35.1	34.5	33.3	35.7	35.5	32.8	35.3	32.6	33.1	36.8	34.5	34.1	32.7	35
32.3	34.1	32.6	31.5	34.3	33.9	31.3	33.3	30.7	31.7	35.1	33.4	32	31.8	33.9
33.3	34.5	34	31.8	33.7	34.7	32.3	33.5	30.7	32.5	35	34.2	32.6	32.1	33.6
31.3	33.8	33.2	29.4	34.7	32.2	31.5	33.6	29.2	30.4	34.8	32.6	31.4	30.1	31.6
31.5	33.3	32.6	29.9	34.7	32.8	32	34.4	29.8	30.8	35	32.8	31.8	30.1	32.4
32.1	33.8	33.2	31.7	35	35.4	31.3	34	30.8	31.3	35.5	33.5	32.2	31.8	34.5
33.7	34.5	33.2	32.2	35.9	35.8	32.5	34.7	30.9	32.4	35.5	33.9	32.4	31.6	33.9
30.7	32.7	32.5	29.3	33.1	31.7	30.2	33	29.1	29.3	33.2	31	30.3	28.7	31.1
32.2	35.2	33.6	31.2	35.3	33.4	32.1	34.2	30.5	31.5	35.7	33.2	33.2	31.4	34
30.2	32.7	31.6	28.4	33.5	31.8	31.2	32.2	30	28.9	33.9	32.4	30.6	29.6	31.3
32.1	33.8	33.9	31.3	34.8	34.1	31.9	33.8	31.3	30.9	34.9	32.8	31.4	31	32.6
31.6	33.3	31.5	30.8	33.4	33.3	31.6	33.2	30.3	30.8	34.7	32.5	31	30.6	33.2
31.5	33.9	33.2	31.1	34.2	33.6	31.3	33.3	30.6	31.4	35.3	32.6	31.4	30.1	31.8
31.8	34.1	32.6	30.6	34.7	33	32.2	34.7	29.5	30.4	35	33.2	31.9	30.2	33.7

30.1	29	27.8	27.2	30.2	26.9	28.1	29.5	27.6	27.8	28.5	29.5	29.2	29.2	26.8
29.7	28.9	27.5	27.5	29	27.1	28.1	29.8	27.9	28	28.9	29	29.4	29.7	26.6
29.7	32.5	27.9	28	31.4	27.7	28.9	33.2	28.5	29.5	29.9	31.4	33	30.3	27.6
30.1	31.7	27.7	27.8	31.1	27.9	29.1	31.1	28.4	29.1	29.3	30.6	32.6	30	27.6
28.6	29	26.5	26.9	29	27.5	28.1	29.9	27.1	28.6	28.5	29.7	30.8	29.7	26.8
30.6	33	29	28.9	31.9	28.5	29.8	33	30.2	30.1	30.6	31	32	30.8	28
30.2	32.1	28.7	27.8	32.8	28.4	29.5	31.6	29.4	29.9	29.4	29.7	31.7	30.5	27.9
30.5	32.1	28.3	27.9	30.3	28.7	30.5	31.1	28.6	27.9	28.6	29.8	30.3	29.6	27.1
29.6	29.2	26.2	26.7	28.8	26.5	27	29.2	26.9	28.2	28.3	29.4	30.5	29.2	26.4
29.9	31.3	28.2	28	30.9	27.8	28.4	30.4	28.3	29.4	29.1	30.4	32	30	27.3
30.5	32.5	28.7	29.1	32.1	28.6	29.4	33.2	29.4	30.1	30.6	31.8	33.5	30.5	28.8
32.7	30.3	29.1	30.1	32.2	29	30.1	31.3	29	29.8	31.2	30.9	31.1	31.3	28.2
29.7	30.1	27.2	27	29.6	27.8	28.6	30.3	27.8	27.4	28.6	30	29.9	29.4	26.6
30.3	32.5	28.7	28.1	31.3	27.8	29.1	33.2	29.1	28.4	29.1	31.3	31.9	28.9	27
31.7	31	29.6	29.1	32.5	28.9	30.2	31.8	28.8	29	30	30.6	30.5	31.5	29
30.2	29.9	26.7	26.9	29.5	27.2	28.1	29.8	27.2	28.1	28.3	29.6	30.5	29.9	26.5
30.1	31	28.4	28.8	30.9	28	29.3	31.9	28.6	29.6	29.7	32	32.5	29.6	27.9
30.4	31.1	28.2	28	30.9	27.9	29	30.5	28.9	29.1	29.4	29.9	30.4	29.7	27.3
30.3	31.7	28.8	28.7	32.6	29.4	29.5	31.4	29	30	30.1	31.6	32.8	31.4	28.1
30.8	33.4	29.4	29	33.1	28.4	29.9	33.2	30	29.9	30.2	31.4	32.7	29.6	28.3
31.6	33.3	30.2	28.6	33.4	30.4	30.9	32.5	30.9	31.1	31.1	31.3	29.5	30.8	28.7
29.8	31.2	27.6	27.4	31.6	28.4	29.8	30.6	28.7	28.3	29.6	30.5	31.3	29.3	27.4
31.6	33.2	29.9	29.3	34.7	30.2	31.1	32.8	30.8	30.9	30.8	30.9	31.6	30.7	29.7
30.7	29.4	29.4	31.5	30.1	29.7	29.7	28.1	31.2	33.6	28.2	29	32.5	32.8	32
30.5	30.1	29.5	31.9	29	29.6	28.6	29	31.4	34.3	29.9	29.4	31.4	31.2	32.2
29.8	28.9	30.1	29.7	29.5	30.4	26	27.3	27.7	30	26.8	26.8	32.1	32.3	30.1
27.1	26	26.8	26.3	26	26.8	24.7	24.5	25.9	27.4	23.9	24.6	28.8	29	27
26.6	26	25.9	28.2	25.5	25.1	25.4	22.6	27.6	32.7	24.4	25.8	26.4	27	27
27.1	27.7	27.8	30	28	25.5	26	24.6	29.3	33.6	24.9	27.5	28.9	30.1	29
28.6	27.8	28.1	28.4	27.1	27.9	25.4	26.2	26.9	28.3	25.7	26.3	30.9	31.2	29.1
26.1	25.7	26.6	28.7	26.4	25.8	24.7	23.8	26.7	32.5	24.6	26.4	27.1	27.9	27.9
27.8	25.6	27	27	26.2	27.5	25.1	25.6	25.8	29.2	24.5	24.7	29.6	29.3	28.4
26.6	25.6	26	25.9	25.1	26.1	23.7	23.6	24.7	26.8	23.5	24	28.3	27.7	27.3
28.8	28	27.1	29.6	26.7	26.5	24.7	24.7	28.9	31.2	24.1	25.4	28	29.5	29.4
25.5	26	26.4	28.5	26.7	25.1	25.3	22.8	26.9	31.8	24.1	26	26.4	27.3	27.3
30.3	28.8	28.9	29.4	29.3	29.7	26.7	27	28.2	30.9	27	27.2	31.6	31.5	30.2
31.9	31.6	30.6	32.2	31.5	31.7	29.1	28.8	32	34.2	28.8	30	34.7	34.7	32.9
31.3	29.8	29.9	31.2	29.8	30.9	28.5	28.3	29.9	33.5	28.6	29.2	32.8	33.4	32.2
29.9	29.5	30	31.8	29.5	29.9	30.1	27.4	31.8	35.5	28.2	29.7	31.4	30.9	32.6
33	32	32.6	33.4	32.4	32.2	31	30.1	33.2	35.9	30.4	30.9	34.8	34.4	34.1
35	34.5	34.4	36.3	35	35.7	32.8	32.7	35.1	38	33.1	32.8	37.4	37.8	37
30.8	29.8	30.9	31.9	30.3	31	28	28.5	30.1	34	28.9	28.7	32.3	32.6	32.1
31.8	31.6	31.7	32.7	31.5	32.1	30.1	29.5	32.1	35	29.3	30.6	33.7	33.2	33.7
28.6	27.6	28.2	29.2	28.1	29.3	26	27.2	28.4	30	26.6	26.6	29.9	30.4	29.2
26.2	24.3	24.4	26	25.2	26.1	22.7	22.5	25.4	27.7	22.8	22.4	27.5	27	26
27.5	27.7	27.6	29	26.8	26.4	27.6	25.2	29.5	32.3	25.5	26.9	27.3	28.8	29.3
31.1	28	28.5	30.9	29.5	30.1	27	27.7	29	32.4	27.1	26.4	32.3	31.5	29.3

30.7	28.4	28.8	31.8	29.7	30.9	28.6	27.8	29.5	31.9	28	27	32.4	32	30
30	29.1	28.9	31.2	29.5	29.6	26.7	28.7	29.3	30.6	26.7	27.9	30.1	31.8	29.8
32.1	31.5	31.8	33.9	32	31.1	31	29.4	33.6	36.5	28.6	29.3	32.3	32	33.6
29.8	27.4	26.6	29.1	28.5	29.3	27.3	26	27.8	30	25.3	25.1	31.8	30.9	29.3
29.2	28.1	29.1	31	29.5	29.1	27.9	25.7	31.5	35.5	26.3	27.8	29	29.6	30.9
30.7	31.6	29.7	28.9	32.6	28.9	28.8	31.8	29.4	29	30.3	29.5	32.1	28.8	28.9
31.1	31.3	30.1	28.6	32.1	29	29.6	32.8	30.1	30.1	31.9	30.3	32.7	30.6	30.2
28.7	29.1	26.7	27	29	27	27.9	30.3	27.4	28	28.7	30	31.3	29.5	27.3
30.5	31.4	28.3	28.4	32.5	28.1	28.5	32.7	29.3	28.7	30.1	30.8	31.5	29.6	28.6
29.3	29.7	27.2	27.5	29.3	27.2	27.9	30.4	27.9	28.3	28	29.4	31.2	29	26.9
28.7	29.4	27.2	27.7	29.4	27.7	28.2	30.7	27.5	28.2	28.7	30.5	31.2	29.8	27.5
31.1	30.9	29.7	30.1	32.8	29.8	30	32.7	30.1	29.8	31.8	31.6	32.4	30.5	30.5
31.4	32	29.9	29.6	32	29.3	29.6	33	28.7	29.6	30.9	31.5	32.1	30.4	29.7
29.7	31.3	28.8	28.2	32.5	28.3	29.1	32.2	28.7	29	30.8	30.1	32.8	29.6	28.6
30.6	33.4	32.8	30.1	33.3	31.5	31	33.5	31.1	30.9	32.8	31.8	32.3	30.4	31.4
29	30.4	27.2	26.7	30.2	26	27.1	31.1	26.9	27.5	29	31	31.6	28.8	26.9
29.6	29.6	26	27.7	29.8	26.7	28.5	32.3	27.9	28.4	29.3	30.6	32	29.1	27.1
30.6	31.2	29.7	27.9	32	29.2	29	32.1	28.7	30.5	31.3	31.2	33.1	29.6	29.4
32.5	31.6	30.5	29.2	33	29.2	29.6	32.9	29.4	29.4	30.9	31.1	31.3	29.3	29.3
29.6	31.5	29.4	28	32.2	29.3	29.7	32.2	29.6	29.7	31.3	30.1	32.7	29.3	28.9
29.8	30.6	27.6	29	29.4	27.7	28.9	31.2	28.3	29.9	29.5	30.1	32.6	29.9	28.2
31.6	32	31.5	31.1	34.5	30.4	30.7	33.6	30.5	30.8	32.9	31.3	32	30.2	30.2
29.1	30.3	28	28	30.3	27.3	28.3	32	27.8	28.3	29.1	30.8	31.3	29	27.4
31	31.9	32.1	29.4	33	30.6	30	33	30	29.8	32.3	31.1	32.7	30.4	30.5
30.4	30.9	28.8	27.8	32	29	29.1	31.3	29.3	28.6	29.7	28.5	31.4	29	28.8
29	30.3	30.4	28.4	33	29.4	29.5	31.5	29.3	30	31.2	29.8	29.9	28.2	28.4
32	32.6	33.3	31.3	34.7	31.1	31.4	33.5	31.6	31.7	33.5	31.6	33	30.8	31
31.2	31.6	29.3	29	33.4	28.4	29.3	33.1	29.2	29.3	30.9	31.1	31.5	29.9	29.3
29.6	30.6	28.5	27.4	32.1	27.6	28.4	32	28.1	28.2	30.1	30.4	32	29.1	27.5
31.6	32.8	33.5	30.5	34.9	31.5	31	33.2	30.5	31.4	33	31.5	32.3	30.2	31
29.2	31	28.8	27.3	31.8	27.8	28.6	32	28.1	28.3	30.2	30	32.7	29	27.8
31.1	33.1	29.5	29.5	32.6	29	29.6	32.7	29.3	29	31.5	32.5	33.6	31	29.1
30.7	31.6	28.6	28.7	31.5	27.8	29	32.7	28.9	29.9	30.3	31.5	33	30	28
29.5	31.6	29.3	27.9	32.3	28.6	29.3	32.4	28.4	28.9	31.3	30.1	32	28.9	28
30.6	32.6	28.9	28.7	32.1	27.9	29	32.7	28.4	29.5	30.1	32.3	33.1	30.6	28.3
30.1	30.9	29.1	29	32.5	28.5	28.7	33.2	29	29.8	31	31.2	32.5	29.8	28.6
30.6	31.3	29.4	29	33.1	28.4	29	32.2	28.9	28.8	30.2	30.2	32.2	29.5	27.9
29.8	31	28.1	27.5	30.7	27.6	28.1	31.5	27.5	28	28.5	29.6	31.1	28.9	27.1
30.6	30.9	28	28.4	31.1	28.2	29.7	32.5	28	28.1	30	32.1	32.8	29.5	27.4
31.1	32.7	29.1	29.4	32.4	28.9	29	33	29.1	29.8	30.7	32.3	31.8	30	28.8
32.3	32.2	32.1	29.9	33.8	31.2	31	33.5	31.8	30.9	33.1	30.9	32.8	30.6	30.9
27.4	28.8	26.6	26.5	30.3	25.7	26.5	29	27.1	28.6	28.7	28.1	29.6	27.8	26.3
28.6	30.2	28.6	26.8	30.5	28.5	28.8	31	28.3	28.5	30	29.7	31.6	28.9	29
29.6	30.9	30.9	28.3	32.2	29.9	30.3	32.9	29.7	31.3	31	30.6	32.8	29.4	30
30.7	32.2	30.6	28.5	33.3	31.3	31.1	33.5	30	30.5	31.9	31.7	33	29.3	30.6
30.6	32.5	30.4	29	33.6	30.8	31.7	33.4	30	30.3	31.6	31.1	34.1	30.9	30.6
27.8	31.1	27.8	26.8	30.5	26.9	29.1	31	26.7	28.2	29.9	29.3	31.7	29.3	28.4

28	28.4	26.1	26.2	27.5	26.9	27.7	29.8	26.9	28	27.9	28.1	30.2	27.2	26.2
28.2	29.8	28.3	26.7	30	27.9	28.6	31.4	27.5	28.4	30	28.8	31.4	28.9	27.9
27.8	29.8	27.7	27.9	30.6	27.6	29.2	31.3	27.3	28.9	28.8	29	30.6	28.4	26.5
30.4	32.4	30.4	28.5	32.1	30.5	31	33.5	29.5	30.8	31.7	31.4	33.5	31.7	30.5
29.6	31.4	31	27.9	32.7	29.8	30.5	32.7	29.4	30.1	32.1	32.1	32	29.4	30.1
29.5	30.3	29.5	27.6	32.2	28.7	29.2	32	29.1	29.6	31	29.3	31.7	28.9	28.7
28.1	30	27.7	28	29.9	27.1	28.2	30.9	27	28.4	28.5	29.1	30.8	28.1	26.9
29	31.3	28.5	27.7	31.1	27.7	28.8	31.2	28.1	29.7	30	29.9	31.8	29	27.8
30.9	32.8	31.8	29.6	33.5	31.5	31.7	34.4	31.3	31.6	33.1	32.5	33.3	30.8	31.1
27.6	29.8	27.5	26.3	30	27.2	28.1	30.5	26.4	27.9	29.2	28.9	30.3	27.6	27.1
27.9	30.5	27.5	26.3	29	27.3	27.8	30.6	26.5	27.3	28.4	27.8	30.4	28	27.6
30.9	32.3	32.1	29	33.7	31.4	30.6	33.4	30.2	31.2	32.7	31.3	33.1	30.8	30.5
30.9	32.5	31.3	29.2	33.3	30.6	31	33.3	30.7	30.7	32.5	31.3	32.5	30.5	29.8
29.7	31.2	29.8	27.6	31.7	29.4	29.4	32.4	28.6	29.7	30.5	29.6	31.3	28.8	28.5
28.6	31.7	28.7	28.4	30.7	28.6	29.7	32.8	28.2	29.5	29.7	29.2	31.7	29.5	28.1
30.7	32.7	31.3	29	33	31.5	31.5	34.4	30.3	30.5	31.9	31.1	32.1	29.9	30.2
28.6	31	28	28	30.4	28.6	29	31.4	27.3	28.8	29.4	28.8	31.4	29	27.4
29	31.4	28.2	27.5	32.2	28.5	29	32.2	28.8	29.9	30	29.5	32.5	29.4	27.4
29.7	31.5	29.7	28	32.4	29.9	30	33.4	28.4	29.2	30.4	30.2	31.8	28.5	29
29.1	31.4	28.7	27.6	30.9	28.8	29.4	32.1	28.2	29.3	30.6	29.7	31.7	30.4	28.6
28.8	30.5	28.3	28.3	30.2	28.9	29.2	32	27.9	30.4	30	29.8	32.9	29.7	28.2
28.6	30.6	28.5	28	31.4	28	28.8	31.1	27.8	30.3	29.8	29.8	32.1	28.9	27.2
28.4	30	27.5	27.2	30.4	26.3	27.2	30.8	27.8	28.1	28.8	28.7	31	27.3	25.9
31	32.4	32.4	29.9	33.5	31.2	31.4	33.3	30.7	31.2	33.2	31.5	32.5	30.5	31.2
30.7	32.2	31.2	28.9	33.2	30.9	30.9	33.2	29.9	30.6	32	31.1	32.3	29.5	30.1
28.5	29.9	27	27.3	29.4	27.1	27.8	31.1	27.2	27.9	28.6	28.9	30.2	27.6	25.7
29.3	31.1	29.8	28	31.6	29.5	30.1	32.8	28.7	29.6	30.9	30.3	32.2	28.8	28.7
30.6	32.6	31.3	28.9	32.7	30.4	30.9	33.8	29.6	30	31.5	30.9	32.4	29.8	29.7
28.5	31.5	29.2	27.7	31.8	28.8	28.9	31.8	27.8	29.1	29.7	28.8	31.5	28.9	27.8
29	31.5	29.2	27.6	31.7	29.1	28.9	31.9	28.4	28.8	31.4	30.3	31.3	28.6	28
32.9	35.4	36.5	33.5	37.3	32.9	33.3	35.7	34.5	34.2	34.8	32.4	33.3	31.4	33.2
33.2	35.2	36.5	33.7	37.9	33.8	34.1	35.6	33.2	34.3	34.9	33.4	34.3	32.3	32.6
32.5	33.7	33.8	31.9	38.2	31.8	32.7	35.1	33.6	33.8	34.4	32.8	33.5	33	33
32.5	32.9	34.4	30.4	36	31.6	32.3	34.5	32.6	32.9	33.3	33	32.6	32.9	32.3
32	34	35.4	32	37.2	31.5	32.6	34.1	33.9	33.7	33.7	31.1	32.1	30.5	31.8
34.1	34.7	35.7	32.3	35.3	31.9	32.4	34.8	33.5	33.9	34	34	32.6	33.1	31.9
34.3	36.3	36.6	34.7	39.1	34.2	33.8	35.9	35.5	35.4	36.7	33.6	34.5	32.9	33.6
32.2	33.8	35	33	36.9	33	33.2	35.7	34.2	34.2	34.3	32.5	34.1	32	31.8
32.9	34.6	36.5	33.7	39.2	34.8	33.9	36	34.8	35.5	34.7	32.8	32.9	32.5	32.8
32.7	35.5	34.6	32.6	39.2	31.8	33	36.3	33.7	34.1	34.1	32.5	33	31.9	32.7
32	32.2	31.7	28.6	35.9	29.9	30.5	32.5	30.8	30.6	32	30.9	31.7	31	30
33.9	35.8	35.2	31.6	38.9	32.2	33.2	36.5	34.3	34.6	34.5	33.6	33.3	33.7	34.3
32.4	35.4	34.9	30.7	38.4	30.9	32.7	35.1	33.8	33.4	33.8	33.2	32.3	32.2	32.8
31.9	31.7	33.3	28.6	37.2	30.2	30.9	32.5	30.1	31.4	31.7	30.9	30.5	30.1	29.8
34.4	34.6	34.7	30.3	37.3	32.2	33.1	36	34.3	34.5	34.7	32.6	33	32.1	32.9
34.4	35.8	36	32.7	39.4	32.5	33.2	36.2	34.4	34.7	34.8	33.5	32.8	31.3	32.7
31.9	32.9	34.9	30.8	38.8	32.5	32.5	32.6	32.8	33.3	34.1	31.1	31.3	30.1	31.9

32.2	32.8	32.8	29.7	37.1	31.4	31.8	33.8	31.4	31.9	32.2	31.8	31.8	31.8	30.2
32	33	33.2	29.9	38	30.6	32.1	34.5	32	32.7	33	32.9	31.3	31.8	31.7
32.8	34.8	36.3	33.3	39	35.6	34	35.6	34.6	33.7	35	31.8	33.6	32	33.8
32.3	33.4	36.3	31.9	39.1	33.7	33.1	33.8	33.5	33.7	34.1	32	32.6	31.1	32.3
32.5	33.8	35.4	32.4	37.4	33.1	32	34.7	33.6	32.6	34.2	31.1	32.2	30.6	31.2
34.2	34.7	37.1	33.4	40.3	33	34.2	35.5	34.7	34.6	34.5	32.5	33	31.5	33
32.4	32.4	33.2	29.6	37.3	31	31.9	33.4	31.2	32.1	32.6	31.2	31.1	31.4	30.9
33.4	36.1	38.2	33.8	38.9	35.2	32.8	36.7	35.8	35.5	35.7	32.7	34	32.4	33
32.5	34.8	36.8	32.1	39.1	34.2	33.8	35.3	34.2	33.5	35.2	32.1	32	31.8	32.6
31.4	33.1	35.4	31.1	39	32.8	33.4	33.2	32.6	33.1	34.5	31.8	32	30.7	32
31.6	32	34.8	30	39.7	30.8	31.5	34	31.3	32.2	31.3	31.3	30.5	30.6	30.7
33.3	36	37.8	33.5	39.9	34.8	34	36.3	35.5	35.2	35.8	32.6	33	31.4	33.1
32.3	32.8	32.2	29.9	37	30.9	31.6	33.7	32.1	32	33.1	33	31.9	31.9	30.9
32.3	32.6	34.4	30.4	38.7	32	32.5	34.2	31.7	33.4	33.3	32.2	31	30.8	30.9
28.6	31.1	29.7	27.1	31.3	29.3	30.8	31.6	28.8	29.4	31.3	31.1	33.4	30.7	30.4
28.5	30.3	28.9	26.8	31.1	29.1	29.6	32.2	28.2	29.1	30.8	30.7	32.2	29.2	28.7
29.2	31.4	30.5	28.1	31.9	29.9	30.5	32.6	29.2	30	31.9	31.4	32.5	29.8	30.5
29.6	31.7	30.1	28.6	32.3	30.2	30.8	33.2	29.2	30.2	32.1	31.8	33.1	30.4	31.3
26.8	30.5	29.5	25.7	30.4	27.9	27.8	29	27.1	28.6	31.2	31	30.2	28.3	29.3
29.6	31.4	31.1	28.5	32.6	29.8	30.2	32.5	29.5	30.3	31.8	30.4	31.3	29.1	29.8
30.2	32.4	31.1	28	33.4	31.6	31.8	32.6	29.3	30.1	33.1	32.6	32	30.2	31
28.4	31.2	29.3	27.2	31.1	29.3	29.8	31.7	28.7	29.1	31.5	30.8	32.1	30.1	30.4
29.8	32.4	31.7	28.6	33.4	31	31.2	32.6	30.2	31.6	33.9	31.8	31.6	30.4	31.2
28.5	31	29.9	28	32.8	29.5	30	31.6	29.5	29.7	32	31.1	32.1	29.4	29.9
31.1	34	33	30.5	35	32	31.7	34.2	30.8	32.3	34.8	32.2	33.1	31.6	32.8
30.2	32.2	31.9	28.8	33.4	30.8	31.7	32.6	31	31.3	33.2	30.9	31.7	29.7	30.6
29.3	32.4	30.8	28	32.7	29.9	30	32.8	29.2	29.9	32.2	30.8	31.8	29.1	29.8
31	33.4	31.5	31.7	33.9	32.8	31.5	32.7	30.4	30.7	32.9	33	31.8	31.7	32
32.2	33.8	33.7	32.2	34.3	34	33.5	33.2	32.5	32	33.5	32.7	32.2	31.9	32.8
31.1	32.9	32.5	31.9	34	33.3	32.6	32.7	30.6	31.5	32.9	32.6	31.2	30.8	32.6
32.1	33	32.4	32	34	33.9	33.2	33.4	32	32.1	33.2	33.4	32.4	32	32.5
31.5	33.4	33.1	32.1	34.2	33.8	33.1	33.7	32.9	32.3	33.9	32.8	31.9	31.9	33.5
32	33.3	32.5	32.2	34	33.6	32.7	32.8	31.4	31.2	33.6	32.7	32.2	31.8	32.7
31.9	33.4	32.4	32.5	33.7	33	32.8	33.3	31.8	31.9	32.7	32.3	31.5	31.5	32.2
31.8	33.5	33.3	31.5	34.3	32.7	33	33.3	31.7	32.4	32.8	32	31.4	32.2	32.5
32.4	33.2	35	33	36.9	34.7	33.8	33.7	32.5	33.4	34.9	33.8	33.1	31.5	34.1
33	34	35.8	32.9	36.6	34.6	33.8	34.7	32	33.8	35.5	33.8	33	31.6	33.6
32.1	36	37.4	32.1	37	33	33.7	32.9	32.7	33.3	34.2	34	35.6	31.3	33.7
31.4	32.2	31.9	31.2	33.3	32.7	32	32.6	31	31.5	32.1	32.2	30.5	31	31.1
32.3	33.7	34.2	31.7	34.7	33.3	33.3	33	31.7	31.8	32.3	31.6	31.1	30.6	32.2
32	34.4	34	32.5	36.2	34.2	33.9	34.2	32.8	33.5	34.6	32.5	31.9	32	33.3
31.6	33.2	34.5	32.2	36.2	32.8	33	34.2	31.9	32.2	33.2	32.4	32.4	31	33.2
33.9	34.8	35.8	33.5	36.2	35.2	34.4	35.2	33.4	32.8	35.2	34.6	34.6	32.2	34.1
33.1	34.5	34.2	33.1	35.4	34.7	34.3	34.3	32.9	33.1	34.8	33.4	33.2	31.6	33.2
32.2	33.7	34.6	32.7	35.7	34.1	34.2	33.7	32.5	33.3	34.2	32.8	31.7	31.4	32.5
26	27.4	26.3	27.3	27.8	28	27.1	28	26.1	26.9	25.1	26.3	28.1	27.7	27.5
26.6	27.8	27.1	28.1	28.6	27.8	28.4	29.2	27.1	27.6	25.2	27.9	28.1	27.3	27.1

27	28	27.8	28.4	28.3	28.1	28.4	28.9	27.1	27.5	26.7	28.5	29.1	27.8	28.4
25.7	27.7	26.9	28.2	28.8	27.8	28.5	28.7	26.7	28.2	27.3	29	29.2	27.9	28
27.2	28.4	27.5	28.8	28.9	28.5	28.9	29.9	28	28.9	26.4	28.7	28.1	28.5	28.3
25.5	26.4	25.2	26.6	27.2	26	25.6	27.6	25.2	25.9	24	25.4	26.1	25.1	25.1
27.4	28.1	27.5	28.3	28.5	27.6	28.8	29.7	27.1	28.3	26.8	28.6	28.9	27.3	27
27.6	28.4	27.5	29.3	29.5	29.2	28.9	29.3	27.5	28.2	26.6	27.5	28.4	27.4	27.3
26.6	28	27.4	28.6	28.8	27.8	28.1	29.4	26.6	27.2	25	27.4	28.3	26.4	26.7
28.4	29.1	29.3	29.4	29.4	28.8	29.4	29.4	26.7	29.8	26.5	28.9	29	28	28.3
27.8	29.1	28.6	29.4	29.9	28.6	29.1	29.4	28.7	28.8	27.2	28.2	29.2	28	27.5
28.1	29.2	28.6	30.2	29.9	28.9	29.8	30.2	30	30.5	27.2	29.8	30.1	28.8	27.5
28.3	30.4	27.3	27.8	29.6	28.7	28.9	30.6	27.4	29.2	30.8	31.5	31.3	28.2	29.3
28.5	30.7	28.5	28	30.7	29.2	29.9	31.8	28.2	29.3	30.7	31.2	31.3	28.6	29.9
28.4	31.2	28.7	28.1	31.8	29.4	30	32.3	28.9	30	32.1	31.9	32.9	29.4	30.3
27.5	30.5	28.4	28.3	31.5	29.2	30.4	31.9	28.9	30.2	31.7	31.9	33.2	29.4	30.3
28.9	31.2	28.3	28.3	31.1	29.5	29.6	31.1	28.5	29.8	31.3	32.1	31.6	29.7	30.6
29.5	30.8	28.9	27.8	29.7	29.8	29.6	31.7	28.2	29.6	31.2	31.5	30.8	29.2	30.8
28.6	31.9	28.6	28.4	31.4	30	29.7	32.4	28.7	30.3	31.7	32.8	32.8	29.5	30.6
29.9	31.7	29.7	28.7	31.4	29.6	30.3	31.8	28.3	30	31.6	32.4	32.2	29.9	30.9
28.5	32.4	28.3	28	31	29.4	29.2	32.1	28.7	29.6	30.7	32.5	33.2	29	29.5
30.9	32.9	31	31.1	33.1	31.9	32	34.2	30.7	32.2	33.6	33.9	34.5	31.2	32.4
29.1	31.1	28.4	28	30.8	29.4	29.9	31.2	28.4	30.1	31.8	32.9	32.1	29	30.4
29.5	31.1	28.7	28.8	32.2	30.6	30.9	32.2	29.2	30.9	32.5	32.8	32.5	30	31.4
29.5	31.8	28.7	28.6	31.4	30.8	30.7	33	29	30.9	31.9	33.3	33.4	29.9	31.4
29.4	31	29.1	28.2	31.4	27.8	30.4	31.3	28.1	30.1	31.5	32.3	31.9	28.1	29.8
31.3	33.3	30.4	29.3	32.5	31.7	31	33.2	29.6	31.2	33.1	32.7	32.6	30.8	32.3
25.7	28.2	26	25.5	28.1	26.6	27.8	28.6	25.6	26.6	27.6	29	30.2	26.9	26.8
23.9	26.2	26	25.9	24.2	25.5	26	26.1	26	27	23.5	25.4	26.2	26.1	26
24.3	26.5	26.1	26.9	24.5	25.5	26.3	26.9	26.4	27.5	23.4	24.9	26.4	26.1	26.9
24.8	25.3	25.7	27	24.7	25.9	26.8	26.5	25.8	26.6	23.4	24.4	25.8	25.4	25.4
24.9	26.8	26.8	27.4	26.9	26.1	27.1	27.3	26.3	27.2	25.2	25.4	27.4	26.5	26.5
23.1	24.9	24.4	26	23.1	24.9	25.2	25.4	24.1	25.3	22.4	24.9	25.2	25.6	24.9
24.8	25.3	25.3	27	24.1	26	26.7	25.5	25.2	26	23.4	25.2	26.1	26	25.6
24.6	26.2	26.2	25.7	26.5	25.5	26.7	26.9	26.3	26.7	23.8	25	27	26.2	26.5
21.9	23.9	23.5	23.9	22.8	24.2	23.7	23.8	23.6	24.5	21.8	23	22	22.9	24.7
24.5	26.5	26.4	26.5	26.3	26	26.7	27.3	26.5	27.3	23.5	25.1	26.6	26.7	25.8
25.6	27.3	27.4	27.7	26.8	26.6	27.7	25.2	26.8	27.5	24.5	26.1	27.3	26.8	26.7
24.1	25.3	26.5	26.2	26.3	25.6	25.2	26.3	25.6	25.5	23	24	26.1	25.3	25.6
25.2	28.9	29	27.8	26.7	26.7	29.4	27.1	27.2	27.9	24.7	25.9	28.2	26.9	27
27.1	28.6	26.9	26.7	26.9	26.7	27	29.5	27.1	27.7	27.7	30.4	30.7	28.9	27.1
23.6	23.6	21.4	23.6	22.8	24	23.5	26.7	22.7	22.1	23.6	24.4	25.9	26.2	22.5
27.4	27.5	25.1	26.4	26.7	26.2	26.6	28.4	26	26.3	26.2	27.7	28.4	27	25.8
24.7	25.7	23.1	25.2	25.5	25.2	25.3	28	23.6	24.5	26.2	27.3	27.6	26.4	24
28.1	28.9	26.2	26.6	27.8	27.3	28.4	30.3	27.8	28	28.2	29.3	31.5	27.5	26.7
26.9	28.6	25.6	25.9	26.7	27	26.6	29.9	27	26.3	27.2	29.2	30.8	27.5	27
27.1	27.6	24.6	26.8	27.4	26.2	26.3	29.2	26.4	26	27	27.8	28.9	28.4	25.4
24.9	25.2	22.1	24.5	24.5	23.7	24.2	26.7	23.8	23.5	24.2	25.4	27.1	25.6	24
27.2	28.5	25.8	26.4	27.3	26.4	27.1	29.6	26.4	27	26.6	29.4	30.4	28.6	25.8

24.7	23.1	21.4	23.1	23.6	23.5	23	25.4	23	22.8	23.7	24.4	25	24.2	22.6
25.2	25.2	24.3	25.6	24.7	25.5	25.9	28.7	25	24.3	24.9	27	27.5	27.5	24.9
28.8	29	27.1	26.4	28.7	27.6	27.5	29.1	28.6	27.5	28.6	30.1	31.5	27.8	28.2
27.9	29.4	27.2	26.7	27.8	27.4	27.4	30	27.4	27.4	27.6	29.9	31.2	27.5	26.8
24.3	26.1	24.5	23.3	25.1	24.3	25.4	26.7	24	25.1	25.3	25.6	25.7	24.8	24.5
28.2	30	27.5	27.1	29	28.1	28.8	30.2	28.2	27.7	29	30.1	31.8	28.1	27.5
26	27.4	25.6	26.2	26.2	26.1	26.2	28.7	26.4	26.3	26.4	28.2	29	28	25.8
26	26.7	23.3	25.4	25.7	24.8	25	27.6	24.5	24	25.1	25.6	27.1	27.3	23.8
24.2	24.8	21.9	23.7	24.5	23.7	23.6	26.9	22.8	23	24.6	23.7	26.9	25	22.7
28.8	30.8	27.7	28.3	28.1	28.2	27.5	30.2	28	28	28.3	29.2	31.5	28.2	27.1
27.9	28.8	26.5	27.3	27.4	28.1	27.5	30	27.5	28.3	27.5	29.1	31	28	26.6
27.8	29.2	26.6	26.9	27.6	27	27.6	29.4	27.2	27.2	28.3	29.1	30.5	27.2	26.2
29	29.5	27	26.9	27	26.5	27.7	30.5	27.3	28.4	27.8	31.1	31.5	28.4	27.2
26.2	27.8	25.8	25.4	26.1	25.7	25.8	28.2	25.7	26.3	27.3	29.1	29.5	27.1	25.8
23.7	24.7	23	23.4	23.8	23.7	24	27.1	23.4	22.5	24.8	25.2	26	23.7	21.6
22.8	20.3	19.7	19.1	22.5	22.1	20.5	23.7	21.9	22.3	20.2	22.5	24.6	22.5	22
25.5	24.6	22.4	24	26.4	23.8	24	26	24.6	24.4	24	25.6	29.1	26.8	24.4
27.6	26.5	24.3	26	27.5	26	26.4	28.1	26.3	25.6	26.2	27	29.5	28	26.5
26	26.6	24.6	25.6	28	24.4	25.7	26.3	25.8	25.6	25.8	25.8	28.2	27	24.4
26.6	26.3	23.4	25	26.8	24.9	25.3	26.9	26	25.4	25.2	26.6	28	27.2	25.5
26.3	25.4	24.4	25.8	25.5	24.8	24.6	27.4	24.8	24.5	24.5	26.1	28.5	26.9	24.5
27.8	25.7	23.8	25.8	27.1	25	25.1	27.5	25.5	25.7	25.5	27	29.8	27.8	25.2
27.6	27.2	24.8	25.6	27.6	25.7	25.9	27.2	26.5	24.9	26.4	27.5	28.5	28.8	26.9
31	29.4	25.9	28.3	30.5	28	29.8	30.1	28.5	29	29.3	31	33.1	30.6	28.6
30.9	29.1	26.5	28.3	30.3	26.5	30	30.2	28.1	29.6	28.1	29.9	32.3	30.3	27.1
28.2	26.8	25.1	26	27.9	26.7	26.9	28.6	27.2	26.1	26.4	28.7	30.5	28.9	26
28.3	28.2	25.9	27.3	28.2	29.6	27.6	29	27.3	26.4	26.9	27.6	29.8	29	26.9
27	26.3	22.9	24.5	27.4	25.3	26.3	27.3	26.3	25.2	26	27.4	28.9	28.2	25.9
26.6	25.9	23.8	25.6	25.7	24.4	24.7	26.8	25.5	24.4	25	26	27.8	27.4	24.7
28.6	27.6	25.7	26	27.6	26.1	26.5	28.1	27.4	26.1	26.8	27.6	30.3	29.8	26.5
27.6	27.3	25.3	26.2	28	26.2	26.2	27.2	27.3	25.2	26.1	27.7	29.2	30.5	25.8
29.7	28	26.3	26.6	28.4	26.8	27.7	29.2	27.3	26.6	27.4	28.4	31.2	29.2	27
33.3	30.7	28.1	28.4	30.6	29.3	28.9	30.6	28.8	28.1	28.3	30	32.8	30.9	27.6
29.5	29.3	28.3	29.3	30.6	29.2	28.5	31.3	29	28.8	28.7	28.3	31.4	30.9	27.7
30.9	30.5	27.8	28.1	30.1	28.5	29.4	30.9	28.4	28.2	28.3	29.7	30.2	29.4	26.7
31.1	29.2	26.9	27.7	30.5	27.6	29.2	30.5	27.8	28.6	28.2	28.8	31.7	29.1	26.8
30.2	29.6	26.8	26.8	29.5	27.5	27.9	29.4	27.4	27.3	28.1	29.5	31.6	30	26.6
29.9	28.7	27.4	27.5	29.7	28	28	29.8	28.7	26.9	27.2	29.1	30.7	30.9	27.4
29.1	28.5	25.5	26.6	28.9	26.5	28.5	29.2	26.4	28.1	27.1	28.8	30.2	28.9	26.3
29.3	28.2	26.5	26.2	28	27.8	26.9	28.3	28.3	27.4	27.7	29	30.5	30.5	26.7
31	30.5	27.9	28.7	30.8	28.5	29.1	31.3	27.9	29.4	28.6	29.9	30.9	29.9	27.3
30.3	29.6	26.4	27.6	30.7	28	29.5	30.6	27.4	29.3	28.4	29.5	30.9	29.1	26.8
30.7	30.3	27.1	28.5	30	28.1	29.4	32.3	27.7	28.7	28.2	29.7	30.9	29.4	27.1
31.4	29.9	27	28.1	32.8	28.3	29.1	31.7	28.1	29.3	28.9	29.7	32.2	29.8	27.4
30.9	29.1	26.9	26.4	30	27.5	28.5	30.5	27.5	28.1	27.6	28.9	30.8	28.6	26.2
31.7	29.9	27.8	27.9	30.5	28.4	29	31	28.5	28.2	28.6	29.5	32.2	30.7	27.2
32.3	30	26.8	28.6	31.1	28.5	30.2	31	29	29.5	28.9	30.5	33.4	30.4	26.9

32.7	31.3	28.6	28.5	30.8	29.4	28.9	30.7	28.1	29.6	29.6	31	32.1	30.7	27.1
30.8	31.3	30.4	28.8	34.9	29.5	30.5	31.5	29.1	29.3	30.1	29.8	30.5	29.5	28.7
30.8	31.3	31	29.8	36.8	30.7	30.8	33	30.7	30.6	32.9	30.7	30.3	28.7	30.6
31.7	31.3	30.3	30	33.7	29.5	30.1	33.2	30	30	32	32.2	31.3	30.4	30.5
30.7	31	29.4	29.6	32.5	28.3	28.9	32.5	28.7	28.7	30.4	31.1	31.1	29.7	29.2
31.9	31.8	31.4	30.4	36.8	30.4	31.1	32.3	30.5	31.5	33.6	32.8	30.9	30.8	29.9
30.7	32.1	34	30.5	37.9	32	32.3	32.5	31.2	32.3	33.8	31.3	30.5	29.7	31.7
31.4	31.6	33.4	29.7	37.5	32	32.3	32.7	30.9	32.5	34.1	32.1	30.7	30.2	30.8
30.9	30.8	32.3	29.4	35.4	30.4	30.5	31.1	30.6	31.2	32.7	30.1	28.8	29.2	30.8
31.8	33.3	32.3	30.1	36	31.3	31.2	34.5	31	31.3	33.7	31	31.5	30.2	32.8
32.2	32.2	32.8	30.2	37.9	31.1	31.7	33.4	30.6	31.9	34.4	32.3	30.1	30	30.9
32.5	33.1	33.2	31.4	36.6	32.3	33	34.4	31.7	32.7	35.1	31.9	31.8	31	32.3
30.2	30.2	31.1	28.2	36.4	29	29.9	32	29.3	30.1	30.8	29.9	29.8	28.8	29
32.6	32.6	31.1	30.9	35.9	29.7	30.8	33.3	29.9	30.3	31.7	31.8	33	30.2	30.3
31.1	33.2	29.2	29.3	32.9	28.7	29.2	33.2	29.3	29.3	29.9	30.9	32.4	30.4	28.4
31.8	32.3	29.6	29.1	34.4	28.8	30	32.1	29.8	30.2	30.7	31.4	32.1	30.4	29
31.3	32.9	35.1	31	37.9	32.2	32.2	32.9	31.8	33.5	35	32.2	31.9	30	31.1
31.5	33.8	35.5	31.7	38.4	33	33	33.3	33.5	33.8	35.6	32.6	33.2	31	32.4
32.1	32	34	30.5	38.3	31.7	32.2	33.7	31	32.3	33.7	31.9	32	31.2	31.2
32	33.3	33.8	31.5	38.2	32.7	33.2	34.2	32.7	34.1	35.9	32.2	32	31.1	33
33.1	32.5	33.2	30.1	38.6	31.4	31.6	34.1	31.1	33.5	34.5	32.8	32.2	31.5	31.1
31.8	31.7	30.6	28.8	35.6	29.8	30.3	33.1	30.6	29.6	29.8	30.8	32	30.7	29
32.2	31.9	33	30.5	38.3	31.1	31.8	33.8	30.5	31.4	33.4	32.7	32.7	30.7	30.5
32	33.1	29.6	29.1	33.8	28.9	30.5	33.9	29.8	29.6	30.3	30.9	32.1	29.1	28.8
31.6	31.8	34	30	38.8	31.5	31.8	33.4	31.2	32.5	32.3	31.4	31.2	29.7	30.7
32.8	32.3	32.5	29.8	37.6	30.9	32	34	30.1	31.2	32.3	31.3	31.1	30.1	30.4
32.5	33.2	31.1	30.3	36.5	30	31.5	33.8	31.1	31	32.4	32.3	32.3	30.6	29.8
31.3	32.3	32.2	30.7	33.2	32.3	31.4	32	30.5	30.7	32.9	32.1	31.5	30.4	31.9
31.6	33.6	33.2	31.8	34.2	33.8	32.4	32.9	32.5	32.3	34	32.5	31.2	30.8	31.9
33.4	34.5	34.9	32.8	35.5	34.6	33.6	33.8	33.3	33	35.3	34.2	33.4	31.8	33.6
32.9	33.6	34.1	32.5	35.2	34	33.1	33.2	32.5	32.7	34.3	33.4	32.7	30.9	32.8
32.8	33.8	33.8	31.6	36.1	35.1	33.1	33.3	32.4	32.2	34.6	33.2	31.9	31.3	32.5
31.1	32.2	33.7	31	34.3	33.1	32.8	32.5	32.2	31.7	33.9	32.1	31.2	29.8	32.5
34	36.2	35.6	32.7	36.4	35.5	34	34	33.1	33.7	36	34.9	32.7	32.2	33.8
31.4	32.4	32.6	30.7	35.2	31.7	31.6	32.5	30.6	31.3	34	31.9	31.7	30.4	32.4
32.2	33.3	34.1	31.2	36.3	32.5	32.7	33.3	32	31.4	34.9	32.1	31.6	30.8	32.7
31.6	32.7	31.6	31.1	33.5	33	31.2	32.7	31.4	30.4	32.5	31.5	31	31	31.8
31.1	32.6	32.8	30	34.9	31.6	31.6	32.1	30.3	30.9	33.4	32	30.6	29.3	31
31.7	32.6	32.9	31.6	34.2	33.4	32.4	32.6	32	31.9	33.1	32.7	32.2	31.2	31.9
35	36.3	37.1	34.8	37.9	36.2	35.5	35.3	34.3	34.4	36.1	35.5	35.4	33.6	35
33.2	34.2	34.1	32.1	35.3	34.3	33.6	33.4	33	32.5	34.8	34.1	33.1	31.7	33.3
33.1	33.9	34.8	32.5	35.1	34.1	33.3	33.5	32.5	32.6	34.3	33	33	31	32.9
33.2	35	34.9	33	35.7	34.5	34.1	34.5	33.2	33.1	34.3	33.3	32.9	31.4	34.4
32.3	34.3	34.1	31.7	35	33.7	32.7	33.3	32.5	32.4	34.5	33.5	33	31.1	32.3
31.7	33.1	33.9	31.9	34.5	33.5	32.8	32.9	32.1	31.4	33.4	32.5	31.1	30.3	32
32.2	33.5	33.2	31.6	34.8	33.2	32.7	32.7	32	30.8	33.3	32.9	32.1	29.8	32
31.9	33.5	34.4	31.1	35.7	33.4	33.4	33.6	31.7	32.5	35.5	32	32.7	30.6	33.1

33.1	34	33.6	31.8	34.5	34	33.4	33.8	32.7	32.4	34.4	34.3	33.2	30.7	32.6
33.3	34	34.1	31.4	35.7	33.5	33.2	32.8	32.5	32.2	35	32.8	32	30.5	32.6
34.3	35.1	36.7	34.3	37.6	35.7	34.9	36.1	34.3	34.2	36.6	34.5	33.8	32.5	34
33.4	33.6	34.6	31.7	36.2	34.8	32.9	33	32.6	32.7	35.5	33.8	32.9	31.6	33.1
33.7	34.2	35.3	32.5	36.7	33.9	33.5	34.2	33.8	32.8	35.9	32.9	33.1	32.1	33.8
31.8	32.4	32.2	31.1	33.4	32.8	31.8	32.2	30.6	30.5	33.5	31.8	31.4	30	31.6
33.9	34.8	34.3	33.2	36.9	35.4	34.2	35.4	33.3	33.3	35.2	33.7	33.7	32.8	33.7
32.4	33.5	34.2	31.2	35.4	33	32.5	32.7	32	32	34.6	32.8	31.9	30.8	32.8
33.7	33.9	34.6	32.4	36.1	33.9	33.7	34	32.8	33	35.6	32.7	32.1	31.3	32.5
32.3	33.1	34.3	31.6	35.9	33.9	32.9	33.6	32.2	32.9	35.8	32.3	32	31.5	33
33.8	34.7	36.3	33.4	37	34.7	34.7	35	32.6	34.7	36.8	33.6	33.2	31.1	33.9
33.1	34	34.5	32.2	36.6	33.8	33.9	33.9	32.8	33.1	35.7	32.6	32.2	31.2	32.9
26.6	25.2	25.7	27.2	25.6	26.8	25.8	24	28.4	31.8	25.6	25.5	28.9	29.4	28.8
26.7	24.8	24.5	26.7	25.7	26.5	24.5	24.3	26.1	28.1	23.9	23.7	28.3	28	25.8
29.1	28.6	27.7	31.1	28.8	28	27.4	26	30.7	34.9	25.8	27.6	29.2	30.9	30.6
31.2	29.3	27	29.8	31	29.9	27.4	29.2	31.1	30.8	27.9	26.9	31.2	30.5	29
27.8	27.1	26	28.2	26.5	26.8	25.5	24.1	29.3	32.8	25	26.7	29.6	29.4	28.7
26.4	25	23.9	25.7	25.5	25.1	24.6	24.7	26.2	29.6	23.8	24.2	27	26.2	26.4
25.8	24.2	24.9	26.2	24.3	25	23.9	22.8	25.4	29.3	21.9	23.2	26.7	28.1	25.1
25.3	23.3	23.7	25.8	23	25	23.7	23.7	25.4	29.4	22	22.6	27.1	26.9	25
26.6	25.2	25.6	28.1	25.8	23.8	24.6	22.7	27.8	32.3	22.8	26.3	27.2	28.1	26.7
28	26.1	23.9	26.9	27.4	26.4	25.4	26.1	28.6	30.6	24.5	24.8	29	27.9	27.6
26.9	25	24.8	27.1	24.5	25.6	23.3	22.8	25.7	25.9	21.8	22.9	28.6	28.5	26.4
27.2	26.2	26.1	27.9	26.9	27.4	25.5	26	27	29	24.8	24.4	30.3	29.9	27.5
26	24.2	24	26.9	25.4	25.5	23.5	23.1	24.8	27.6	22.5	22.8	27.2	27.4	26
29.5	27.6	27.8	31	28.4	29	27	26.3	30.4	33	25.5	27.2	28	30.5	29.6
27.6	28.8	26.9	30	31.2	29.8	28	29.6	31.2	30	27.5	27.9	30.8	31.4	29.2
29.2	27.3	26.8	28.8	28.3	27.7	27.7	26.6	30.5	32.6	26.3	26.9	30.2	30.2	28.2
26.7	25.6	26.4	29.2	26	25.4	24.7	23.6	28.9	31.7	23.8	25.9	28.6	29.4	27.9
30.8	28.5	25.6	28.1	29.8	28.2	26.7	29.6	30.5	30.8	26.5	27	31.4	30.4	28
29.5	28.1	28.4	30.1	28.4	28.8	28.6	27.4	30.1	33.5	26.8	27.5	31.5	31.7	29.9
26.4	25	24.5	26.7	25.7	26.4	24.3	25	26	28.9	23.4	23.7	29	28.3	26.7
27.1	26.1	23.4	25.5	26.6	25.2	24	26.5	26.1	27.4	23.6	24	28.5	28.1	25.5
30.3	29.8	26.6	28.8	30.3	29.3	28.4	31.6	30.6	29.8	28.9	28.8	33	32.7	29.7
26.9	25	24.1	26.5	25.7	25.7	24.1	23.5	26.1	28.6	23.2	23.8	28.3	29.2	27.2
28.3	26.7	25.9	28.4	27.5	28.2	26.2	26	27.1	30.5	24.9	24.8	30.4	29.8	28.6
25.3	24.2	23.6	24.7	24.6	25	24	23.2	26.5	28.5	22.5	24.2	26.8	26.5	24.8
26	24.2	23.6	25.7	24.7	24.9	22.4	22.7	24	25.1	21.8	22.5	27.5	26.8	25.3
29	28.9	28.3	28.9	29.4	28.7	27.4	28.7	31	31.5	27	27.1	30.3	30.1	27.8
29.5	28.2	27.4	29.1	29	27.5	27.1	26.7	30.1	31.9	26.1	27	29.7	29.6	28
30.2	29.2	26.4	29.5	30.4	29.7	27.3	28.9	29.9	31.3	26.6	26.5	32.5	30.9	30
26.9	25.7	24.5	25.5	26	25.8	24.7	25.3	27.5	29.4	24.5	24.8	26.9	27.1	26
29.1	29.1	28.3	28.9	31.3	32.2	28.6	30.1	33.5	33.7	28.3	28.7	31.4	31.9	29.3
27.7	25.8	25.1	26.9	25.7	26.3	23.8	24.3	25.1	27.6	23.1	23.5	29.1	28.4	27
32.4	30.8	28.5	30.7	32.6	31.1	29.5	32.6	32.9	32.3	29.8	29	33.2	32	30.4
31	30.5	27.3	30.8	32	30.9	28.5	31.6	32.5	33.2	28.2	28.9	33.7	31.7	29.9
27.8	28.6	25.9	28.9	29.9	29.9	27.4	29.6	31.8	32.8	27.3	28.1	30.5	30.9	28.1

27.8	26.1	25.6	26.9	27.1	27	25.6	25.4	28.3	30.6	25	25.6	28.4	28.5	27.1
29.8	30.1	27.5	28.5	31.5	30.1	27.9	30.9	29.7	31.6	28.4	28.1	32.1	31	29.7
28.1	29.1	26.2	28.5	29.9	29.6	27.1	30.5	30	29.8	26.7	27.1	30.8	31.5	28.4
29.2	28.8	27.2	28.5	29.5	28.8	27.2	26.5	30.9	31.9	26.2	27.2	30.1	30	28.1
31.7	33.1	28.2	30.6	33.1	31.7	29.2	33.2	32	32.8	29.9	30.3	34.3	32.3	31.4
30	30.6	27.1	28.5	31.1	30.1	28.6	32.1	31.1	30.9	26.8	28.8	31.9	31.6	29.6
30.2	31.3	27.3	31	32.1	31	29.7	31.4	31.3	31.8	29.6	29.8	33.1	31.7	30.7
31.2	32.1	28.6	30.5	33.4	32	30.1	32.9	33	33.1	29.2	31.5	34.8	34.3	31.8
29.9	31.1	28.7	30.8	33.5	31.3	29.7	32.8	32.5	33.6	29.2	29.4	32.8	33.3	30.3
24.8	27	26.4	23.1	27.4	25.8	25.5	26.7	24.3	24.6	27.7	27.1	26.6	25.3	26.1
28.9	30.7	28.9	28.6	30.7	30.3	29.6	30.9	28.6	29.7	31.3	31.2	30.4	28.9	31
30.5	34.5	31.2	29.8	33.9	31.3	30.8	33.9	29.8	30.8	35.1	32.7	31.6	30.4	32.2
29.4	30.6	29	29	29.9	29.9	29.6	30.1	28.4	29.6	30.6	30.9	29.5	29.5	29.7
29.9	32.7	30.5	27.6	32	30.8	29.6	31.9	28.4	28.7	33.3	32.4	31.3	29.5	30.8
32.1	34.5	31	30.1	32.2	31.8	29.4	32.3	29.8	30.3	33	33.1	31.9	30.8	32.5
33.9	35.3	32.6	31.5	34.3	33	31.9	33.6	31.5	31.9	35.7	35.2	33.5	32.6	34.7
31.3	34.7	30.5	29.5	31.9	31.2	30.8	33.1	28.1	30.5	34	33.6	32.8	30.8	32.4
29.9	32.8	29.8	28.5	31.5	30.3	30.1	32.5	28.8	28.9	33.4	32	31.1	29.3	30.2
32	34.5	31.7	31.1	33.3	32.1	31.6	33.3	30.9	31.4	35.2	32.7	31.9	31.5	33.2
30	31.9	29.6	28.8	31.6	30.6	30.4	32	28.7	29.7	32.5	32.4	30.3	30.2	31.1
24.1	25.9	24.9	21.8	26.7	25.4	24.6	26.1	23	23.5	26.6	25.4	23.5	22.9	24.9
30.9	33.8	30.3	29	31.6	30.8	30.6	33.3	28.5	30	33.8	32.5	31.7	30.6	32.2
31.7	33.7	30.6	29.8	33.1	31.8	31.4	33.3	29.2	30.4	34.1	33.7	32.8	31.1	32.4
32.6	33.7	31.8	32.5	33.3	33.1	32.5	33.5	31.6	31.7	34.7	33.7	31.2	32.4	33.1
30.8	32.8	31.2	28.3	32.9	31.2	30	32	28.8	29.3	34	32.3	32.3	29.8	31.3
29.8	33.5	30.2	28.6	32	30.6	30.2	32.5	29	28.7	32.7	32.1	30.7	29.5	30.9
27.7	29.1	28	27.5	28.9	28.9	28	28.6	26.7	27.9	30.3	29.5	27.6	28.2	29.4
32.2	33.9	31	30.2	32.6	32.1	31.3	33.2	29.1	30.3	33.4	33.7	32.4	31.5	32.4
30.3	34.4	30.6	29.6	33.6	31.2	31.2	33.6	29.7	30	35	32.9	31.7	30.6	32.2
26.5	28.3	27.5	25.1	29.6	27.7	27.6	27.8	25.6	26.2	29	28.5	27.2	26.2	27.2
31.5	33.6	31.1	30.9	33.2	32.2	31.9	33.5	31.2	31.9	33.8	33.4	32	31.5	33
30.8	33.4	30.3	29	32.9	31.5	30.6	32.8	29.3	30.2	34.2	33.5	31.6	30.1	32
32.2	34	31.4	31.1	33.1	32	31.3	33	30.4	30.8	35	34.2	32.2	31.8	33.4
31.8	34.8	31.3	31.2	33.4	32.8	31.5	33.6	31.1	31.6	33.9	33.9	33.4	30.8	33.6
27.5	31	30.9	26.8	31.9	29.1	29.5	30.1	27.5	28.2	31.7	30.3	30	28.3	29.6
28.7	30.9	29.5	26.9	31.4	30	28.9	31.4	27.4	28.2	31.2	29.8	28.5	27.1	28.5
30	32.7	30.9	28.1	33	31.8	31.3	32.7	29.3	29.6	34.1	31.9	31.5	29.3	30.5
33.1	35	31.3	30.7	33.8	33.7	32.7	34.6	30.6	31.5	35.1	35.5	33	32.4	34
28.5	27.4	25.1	25.8	29.5	28.4	26.1	28.7	27.8	28.8	25.8	27	30.5	31.1	27.7
29.2	28.5	27.4	28	31.1	29.7	28.6	29.5	28.9	30.9	27.5	28.7	30.5	31.8	28.6
30	29.6	27.5	28.1	32.2	30.2	28.4	30.2	30.2	31.7	28.4	28.6	30.2	30.6	28.6
25.9	25.2	23.9	24.8	25.4	25.3	24.6	27.4	25.9	24.6	23.9	24.1	27.4	28.8	24.4
30.2	29.4	26.8	27.2	30.8	29.6	27.9	29.6	29.6	30.1	27.4	27.6	31.5	31.2	28.4
26.7	27.3	25.6	27.4	27	27.7	25.8	27	27.8	26.3	24.5	25	27.9	28.5	24.5
30.5	28.6	26.6	27.7	30.2	30.5	28	29.4	29.3	30.6	26.5	28.1	31	32.2	28.8
28.7	26.8	25	26.7	28.7	27.4	26.9	29	27.5	27.3	27.8	28.1	30.3	30.3	27.3
29.2	28.2	26.9	27.9	28.1	28.3	27.1	29.5	29.9	28.8	26.7	27	29.9	30.7	26.7

28.1	26.8	25.1	25.7	28	26.8	26.8	28.2	25	29	26	27.4	30.5	30.3	26.6
27.7	27.3	26.4	28.6	28.4	28	27.3	28.8	29.4	27.3	25.7	25.7	29	29.1	25.5
27.3	26.9	26.2	27.3	27.6	26.2	26.1	28.8	26.8	25.9	25.9	26.4	29.6	29.2	26
30.1	29.1	27.5	29.7	30.6	30.1	28	30.5	31.3	31.3	26.6	27.6	32.8	33.1	30
29.5	28.8	27.6	27.7	29.1	28.1	27.7	30.1	27.1	27.2	27.3	27.9	30.4	30.2	27.5
29.4	29.6	27.6	28.2	30.2	29.4	29	29.9	29.2	30	27.1	28.2	30.8	31.9	28
27.3	29.1	26.2	28.5	29.6	29.4	26.3	29.2	30.3	29.9	25.8	27	29.8	31.1	26.4
33	31.5	29.8	28.6	31.8	30.2	31.4	31.8	31.5	32.6	30.1	31.9	32.9	33.6	29.9
31.4	29.9	28	28	30	28.2	29.5	30.5	29.1	29	28.6	30.9	32.4	32.3	28
30	30	28.3	27.9	29.9	29	30	30.6	29.4	30.9	27.7	30.1	32.5	31.3	27.5
30.5	29.8	28.8	28.4	29.8	28.5	29.1	30.6	29.1	28.8	28.5	29.6	31.3	31.1	28.5
30.6	28.4	27	27.9	29.9	28.1	28.8	31	28.5	29.8	28.6	29.1	31	30.9	27.2
30.5	28.8	28	28.5	30.5	29.8	29.9	31.5	28.9	30.7	28.5	29.5	32.6	31.2	27.2
30.6	29.1	27.9	28.2	30.5	28.6	28.6	30.2	29.7	29	28.2	28.9	32.1	31.3	27.4
32.4	30.5	27.6	27.1	30.8	28.5	30	31.1	28.9	30.9	27.9	28.6	32.5	31.2	27.7
32.8	32.2	31.8	30.9	33.8	31.8	31.1	32.8	31.8	32.5	31.3	32.8	32.7	33.3	29.8
30.3	29.5	29.1	28.2	33.2	28.2	31	29.7	29.5	30.1	30.7	31.2	29.9	29.8	28.4
30.7	30.6	30.4	29.5	32	29.3	29	30.1	30.4	30.7	30.5	30.4	30.9	31.6	27.8
30.5	30.9	31	28.9	34.3	30	32	31.5	30.7	31.2	31.4	32.6	31.6	31.4	29.5
32.5	33	33.3	30.5	33.7	32.2	31.8	33.3	33.2	31.7	33	33.2	32.8	32.6	30.9
32.3	30.9	30.4	29.3	32.4	30.4	31	31.3	30.5	31.1	31.4	32	31.9	32.8	29.3
31.9	31.5	30.7	30.4	33.1	30.8	32	31.9	31.4	32.5	30.8	32.6	32.2	33.3	30.1
32.7	33.3	32.5	29.3	35.9	30.3	31.8	34.7	33	32.8	33.7	32.6	31.5	30.6	32.1
30.1	29.7	28.8	29.1	31.9	29.1	28.8	30.3	29.2	31	28.2	30.5	31.1	32.4	27.9
33.3	34.2	34.9	31.9	34.7	32	32.3	34.7	33.1	32.7	33.8	33.6	32.8	32.1	31.8
32	31.9	31	30.5	34.6	30.9	32.5	32.2	31.6	33.6	32	33.6	31.6	32.9	29.5
33.4	33.7	34.6	30.8	35.9	31.8	33	34.8	32.5	32.4	33.4	32.9	32.3	32.8	32
31	32	31.1	29.2	34	30.5	30.8	33.3	31.3	30.5	31.1	33.2	30.4	29.7	29.6
31.9	33.2	29.3	28.1	33.5	30.2	30	32.5	30.7	31	30.8	32	30.3	31.2	29.8
30.7	31.9	31.6	27.8	33.9	29	30.7	32.8	30.3	29.5	30.7	31.1	30.5	29.5	30.5
29.8	30.5	30.7	27.4	33	29	29.7	31.1	29.1	29.4	30.4	30.2	28.9	29.1	28.5
32.4	33.7	31.6	30.1	33.7	31.1	31.6	33.7	31.7	31.6	32.4	32.2	31.2	32.5	30.5
30.4	31.1	32	27.8	34.3	29	31.3	32	30.8	31.2	32.1	31.2	29.7	29.3	29.7
31.2	32.1	31.4	28	33.5	29.3	30.9	33	30.2	30.7	31.4	31	30.5	30.2	30
30.3	31.5	30.9	28.6	34.6	30.5	32.2	30.8	31.2	31.7	30	31.7	30.6	31.1	30.7
32.3	33.8	32.6	29.5	35.8	30.2	32.4	34.1	32.4	32.1	33	31.5	30.7	30.1	31.1
31.5	33	31	28.8	34.6	29.8	31.2	33.2	31.8	32.1	31.2	31.8	33.3	30.3	30
30.6	30.7	29.7	28.3	32.6	28.6	29.8	31.4	29.4	29.1	30.3	31.2	30.8	30.2	27.9
31.2	31.9	31.7	28.6	34.1	29.8	31.5	32.1	30.3	30.6	32.3	31.6	30.5	30	30.4
31.1	33.4	31.4	28.9	35.9	30.2	31.1	33.4	31.2	32.3	32	32.2	30.9	31.4	31.1
32.2	32.8	30.9	28.3	35.7	29.8	30.6	33.3	31.3	32.2	31.4	31.8	30.3	31.2	31
31	32.1	32.2	28.5	34.1	29.6	31.8	32.7	30.4	31.7	32.3	32.9	31.9	30.9	30.4
34	35.3	36.9	31.8	38.1	32.3	33.3	35.4	33.5	33.8	34.1	33.8	32.8	33.1	33
31.1	32.7	31.3	29.1	34.4	29.8	31.6	33.3	31.1	31	31.9	32.2	30.6	30	28.8
30.9	32.2	30.2	28.9	33.8	30.1	31.1	32.5	30.6	30.4	31.1	31.6	30.5	30.9	29
32.5	34.4	32.5	30.2	35.2	31.4	33.3	34.8	32.2	32	33	33.4	32	32	30.9
34.7	36.2	36.2	31.1	38.4	31.7	33	35.8	34.6	33.4	34.8	34.1	33.1	32.7	33.7

30.6	31.6	30	28.1	32.6	29	30.2	31.2	29.5	29.7	30.6	31.9	30.5	30.9	28.4
31.5	32.2	31	28.8	33.1	30.2	31	32.3	30.2	29.9	31.1	31.6	29.8	30.9	29.4
31.7	34.1	30.4	28.6	34.4	29.9	31.4	32.7	31.2	30.7	30.5	31.2	30.3	30	28.6
31.1	32.4	31.5	28.6	33.5	29.5	30.7	33.1	31.1	31.1	30.7	31.1	30.2	29.6	29
32	34.3	32.9	30.7	37	31	33.2	35.6	33.4	33.7	33.8	31.9	30.7	30.9	30.4
31.3	33.3	31.5	29.9	37	31.4	32	34.3	31.6	32.2	32.2	32.6	31.6	30.9	31
30.5	30.3	29.7	28	32.3	28.5	29.5	31.4	28.4	28.9	29.8	31.1	29.2	30.2	28.5
32.6	32.4	30.7	29.6	33.8	30.1	31.2	32.5	29.5	29.6	30.8	32	31.4	32	29.9
31.6	32.9	31.4	29.7	33.8	30.1	31.8	33.9	31.2	31.5	32.5	32.8	31.1	31	30.7
32.2	34.7	30.8	30.2	35.7	31.5	32.6	33.5	32	32.5	32.8	32.9	31.4	32.2	30.8
33.2	34.9	32	29.9	36	31.5	31.9	34	32.1	32.2	32.2	33.1	31.1	31.7	30.6
32	34.6	31.5	29.4	36.6	31.6	32.3	34.5	31.6	32.6	31.5	31.8	31	31.6	30.5
31.9	33.9	31.5	30.4	35.9	31.9	32.6	34.3	31.1	32.6	32.3	32.6	32.3	32.5	30.9
32	32.5	31	29.2	34.5	30.1	31.2	32.6	30.9	30.4	31.4	32.7	31	31	29.1
25.6	26.9	27.1	28	28.2	26.9	27.7	29.3	27	27.2	25.5	27.6	28.5	27.1	27.2
26	26.8	27.2	27.4	27.6	26.2	27	28.6	26.4	26.9	24.5	25.5	27.5	26.5	26.4
22	23	23.8	24.8	23	23.9	23.5	23.6	23.4	24	21.9	24.3	24.3	24.5	23.2
24.2	26	26.7	27.4	26	26.3	26.5	26.3	25.7	27	25.3	27.2	27.4	26.9	26
25.9	28.3	29.1	29.4	29	28.3	29.9	29.4	27.4	29.5	27.2	29.8	30.3	29.8	29
26.3	28.2	26.5	27.1	28.8	28.5	29	30.6	26.7	28.2	27.8	30.2	30.5	27.5	28.7
28.4	30.5	28.2	28.5	31.2	29.9	29.8	31.5	28	29.7	29.6	30.6	31.8	28.1	28.9
27.1	29.1	27.2	27.5	29.3	28.7	28.6	29.9	26.6	28	27.3	29.3	30.5	27.8	28.3
26.9	28.9	26.4	27	29.8	28.6	28.1	30.4	26.8	28	28.4	29.4	30.9	27.1	28
28.7	30.1	28.4	28.7	31.4	30.3	29.9	31.6	28.2	29.1	29.1	30.2	31.2	28.4	28.6
27.1	29	26.5	26.6	28.7	28.2	28	29.7	26.6	27.6	27	27.9	28.5	27	27.3
27.8	29.5	28.1	28.2	31.2	30	29.9	31.8	27.7	29.2	28.8	30.2	31.6	28	28.7
29.2	30.8	28.3	29.1	31.5	30.5	30.5	31.8	28.3	30	29.7	30.5	31.9	28.9	29.3
28	30.3	27.3	27.8	29.3	30.4	28.3	31	27.8	29.2	29.4	30	30.6	28.2	28.9
25	26.7	23.7	24.9	25.3	28.2	25.1	28.6	24.8	25.6	27.3	26.8	25.7	25.7	25.6
26.1	29	26.9	27.7	29	28.2	28.4	29.6	26.6	27.7	27.3	28.8	29.9	27.3	27.3
28.4	31	27.8	29.3	31	30.4	30.5	32.5	28.9	30.2	30.6	31.7	32.2	29	29.3
31	32.2	33.6	32.8	34.8	33.3	32.8	32.7	33	33.3	31.2	33.4	33.3	33.7	32.8
31.6	35.7	35	34.8	37.8	34.1	34.8	35.2	33.2	34.3	33.6	34.8	32.9	34.7	33
33.1	31.6	33.4	33.2	34.5	32.2	32.5	32.2	31.8	32.1	29.9	32.2	32.8	33.2	31.6
31.5	32.4	34.6	32.9	35.4	33.2	32.8	32.7	32.5	33.2	31.2	34	33.9	33.7	33
32	35.2	35.2	33.8	36.8	33.9	33.3	34.7	33	34.5	33.9	34.8	32.3	33.2	32.8
22.4	21.9	23.7	22.5	24.6	22.4	22.8	21.9	22.8	21.7	21.4	23.4	22.1	23	20.8
30.8	31.7	34.2	33.5	35.5	33.7	33.7	33.5	31.8	32.9	31.7	33.2	31.2	33.2	30.5
32.7	33.4	36.5	34.7	36.7	34.3	34.2	35.1	32.4	34.5	31.2	34.5	32.9	34	31.9
23	25.4	26.2	26.3	27.5	25.3	25.9	25.8	23.9	25	23.5	25.3	23.9	24.9	22.7
33	34.4	36.5	35.7	37.8	34.9	35.3	35.1	34.4	36	32.7	35	33.5	34.9	32.7
29.4	31.2	33.2	32.2	33.7	31	30.9	31.8	30.6	31.4	29.2	31.4	30.6	31.1	29.7
27.6	27.5	29.5	29.8	30.5	27.8	27.9	27.9	27.3	29.2	26.7	28	27.9	29.3	27.1
31.9	34.2	35.2	34.2	36.9	33.7	32.9	34.4	33.6	34.2	33.1	34.2	32	32.9	31.9
29.8	30.9	34	32.2	33.7	30.3	31.6	31.9	30.4	32.2	28.8	30.7	30.9	32	29.8
32.7	33.9	36.1	35.1	36.8	34.6	36	35.5	34.2	35.1	32	34.7	33.8	35.5	33.2
32.6	35.2	35.5	34.2	37.3	34.1	33.2	35	33.6	35.6	35.1	34.5	31.8	33.7	33

29.9	33.2	33.8	33.4	35.4	32.3	32.3	33.6	32.5	33.5	32.2	32	31.9	32.9	31.6
29.9	29.1	31.2	30	30.9	29.8	29.8	30	30.2	29.6	27.3	29.9	29.7	31.5	28.7
28.4	28.1	30.9	29.3	31.8	29.2	30.4	29.4	30.2	29.3	27.5	29.4	29.2	29.4	28.3
30.9	32	32.5	31.4	35	31.2	31.4	32	31.7	30.9	31.3	31.9	29.4	30.6	28
27.7	28.9	27.7	29	32.9	28.4	28.2	29.6	28.4	29.4	28.5	29	28.1	29.1	27.8
33.5	35.2	36.8	35.6	37.4	35.2	35.3	35.7	34.7	34.9	33.4	35.3	34	35.3	32.8
30.4	33.2	33	33.2	35.3	33.1	32.8	32.9	32.2	32.3	32.1	32.8	31	32.1	30.8
29.9	32.1	32.9	30.9	34.4	31.2	29.8	32.3	31	32	31.5	31.4	30.7	29.7	29.8
32.5	36.8	37	35.5	38.7	34.8	36.6	36.7	34	35.3	34.3	35.1	33.5	35.5	32.5
27	25.2	27.5	27.1	28.5	25.5	25.7	25.7	26.5	25.7	24.6	26.6	26	27.1	23.9
28.7	27.7	29.1	28.1	31.6	27.8	28.5	28.6	26.9	27.5	27.8	28.6	27.5	27.7	25.8
32.5	35	37	36.2	38.3	35.9	36.8	36.1	34.5	35.5	33.1	34.6	33.3	35	33.5
30.6	31	32.4	32.5	34	30.9	31.6	31.2	31.1	30.5	29.4	31.4	30.4	30.6	29.4
31.5	32.6	32	31.8	33	32.8	30.8	29.6	33.4	33.1	31.6	30.1	35	34.1	33.7
32.2	31.5	31.4	33.1	31.9	34	29.8	30.7	31.9	34.3	30.5	30	34.2	33.8	32.8
32	31.9	31.9	32.3	32.2	33.1	30.3	29.1	33.7	33.9	31	30	34.4	33.3	33.1
32.3	32.7	32.6	32.7	33.3	32.5	31.9	31.3	33.3	33.6	30.5	30	34.6	33.3	32.8
34	33.4	33.1	33.6	33.5	34	31.5	31.3	34.3	35	31.8	31.4	36.1	35.7	34.7
33.9	34.4	34	33.8	34.3	35.5	33	32.6	34	34.9	33	32.4	36.4	35.8	35
30.4	30	30.8	30.2	30.3	31.5	27.5	29.1	29.5	30	29.3	28.1	31.8	31.9	30.8
36.5	35.5	35.6	36.3	36.7	37	34.5	33.4	36.5	37.9	34.5	33.7	38.2	38.4	36.6
31.1	29.9	30.2	30.8	30.1	30.9	27.9	28.7	29.5	31	27.4	27.7	31.1	31.3	29.4
33.1	32.3	32.3	32.3	32.2	33.3	29.7	29.8	32.7	33.9	30.3	30.5	34.2	34.4	32.5
28.1	28.4	28.6	28.5	28.7	30.3	27.4	27.1	28.9	29.4	27.3	26.6	30.6	29.9	28.9
35.9	36.6	37	37.6	37.5	37.4	35.2	35.7	34.2	36.6	34.6	34.9	37.2	37.9	36
37.4	38.9	39	39	39.1	39.3	37.3	37.5	36.2	38.9	36.3	36.3	39.6	39.8	38.4
24.2	26.6	25.1	25.4	26.8	26.5	26.6	27.7	25.2	25.6	25.3	27.7	29.3	26.5	26
19.6	21.5	20.5	22.4	21.9	21.3	22.2	22.9	21.3	21.6	20.5	22.7	23.3	23.1	21.9
25.8	29.2	27.8	28.4	29.6	29.2	28.8	29.7	27.3	28.7	27.9	29.6	29.8	28.4	27.8
25.3	26.3	24	26	27	27.2	26.3	26.7	25	25.6	24.9	25.8	26.5	24.9	25
26.2	28.4	26.9	26.9	28.6	27.7	28.1	28.7	25.7	27.8	26.5	28.7	29.5	26.8	27.3
24.9	27.7	26.5	27.6	29	26.9	28	30.2	26.5	27	27.1	29	30.5	27.5	27.6
22.2	24.5	23.8	24.7	26.3	24.5	25.2	26.6	24	24.3	24	25.5	26.6	25	24.2
23.4	26	26.1	26.4	26.5	25.7	27	27.6	25.1	25.4	25.1	27.3	28.4	26.4	26.2
27.6	30.3	26.6	28.5	30	29.2	29.1	30.6	26.4	28	28	29.4	30.7	27.6	28.4
27.1	29.3	27.3	28.1	29.5	28.3	28.6	30.2	27.2	27.7	27.5	28.8	29.9	27.3	27.4
26.5	28.8	27.5	29	28.8	27.2	28.5	29.2	26.3	28	27.3	29.3	30.3	28.1	29.2
24.2	26.7	26.3	27.3	27.2	26.3	27.6	28.2	25.6	25.7	26.1	28.5	29.3	27.4	26.4
27.4	29.2	27.1	27.6	28.1	28.8	28.2	29.1	25.9	26.8	26.8	27.2	27.8	25.9	26.3
24.8	28	26.3	26.7	28.3	26.6	27.8	29.2	26.4	26.9	26.4	28.9	29.9	27.4	27.3
23.7	26.3	25.2	25.6	26.9	25.1	26.2	27.5	25.1	25.2	25.7	27.7	28.6	26.9	25.8
27	29.4	29.4	29.6	29.6	29.5	29.7	30.5	27.9	29.2	28.3	30.4	30.9	29	29.8
23.1	25.9	25.7	26.1	25.5	24.9	25.8	27.4	24.5	25.1	24	26.4	27.7	25.2	24.7
22.9	25.2	24.9	26.1	25.2	24.4	25.4	26.7	24.1	24.5	24.2	25.8	27.2	24.7	25
25.3	26.5	27.5	27.8	27	27.1	28	27.6	27	25.5	24.8	26.6	27.4	27.6	26.4
28.2	29.9	27.5	28	30.5	29.5	29.4	30.6	27.4	28.9	28.9	29.9	30.8	27.6	29.5
24.6	28	28.9	28.7	28.8	29.2	28.7	29.7	26.9	27.4	27.1	29.4	30.3	28.6	28.2

23.9	26.3	26	26.9	25.7	25.6	26.4	27.4	24.7	24.5	24.4	26.4	28.1	26.8	25.3
23.5	25.9	25.8	26.3	26	25.5	26.7	27.6	25	25.8	25.1	26.6	27.4	25.7	26.1
26.1	29.1	27.1	28.2	29.2	28	28.9	29.4	27.5	27.6	26.2	28	29.3	27.6	27.3
26.7	29.6	27.6	29.1	29.7	29.4	28.8	29.9	27.3	28.2	27.3	28.8	29.8	27.4	27.5
23.1	25.9	26.3	26.7	25.2	25.1	25.6	26.6	24.5	24.9	24.8	26.7	27	26.4	24.3
24	26	26.5	27.1	25.9	26.2	26.5	26.8	26.3	25.5	24.3	26.9	27.5	27	26
24.2	26.6	26.7	28	26.2	26.3	27	28.5	26.1	26.1	25.2	27.5	28.6	26.8	26.4
23.4	26.3	26.1	26.4	26.1	26.1	26.7	28	25.6	25.5	24.8	27	28.2	26.2	25.6
23.4	24.2	24.9	25.8	24.6	24.4	25.5	26.1	24.5	24.9	23.3	26.3	26.1	25.6	24.9
23.7	26.2	25.6	26	25.8	25.6	26.2	27.6	24.8	25.5	25	26.9	28.3	26.7	26
25.2	27.1	26.6	27.9	28.5	26.7	27.9	28.4	25.2	26.5	25.9	27.3	29.2	26.9	26.7
25.2	27	27.2	28	27.1	26.8	27.9	29.4	26.5	26.7	25.9	27	28.8	28.1	27.2
24.5	26.7	27.6	28	26.6	26.9	27	27.9	26.7	26.5	25.2	27.7	27.9	27.5	26
23.5	25.9	26.3	27.4	26.5	26.5	26.9	27.7	26.5	26.4	25.2	27.3	27.2	27	26.4
27.2	30	28.7	29.1	30.1	29.8	29.9	30.9	28.1	29.3	28.6	30.1	30.5	28.2	28
23.1	25.6	25	25.7	26.5	26.2	26.5	27.3	24.7	25	24.8	27.2	28.4	27	25.7
25.9	28.7	28.5	28.3	28.5	27.4	28.2	29.9	27.2	27.7	27.4	29	30.1	28.3	27.1
26.2	28.2	28.4	29.2	28.2	27.8	29	29.7	27.7	28.3	26.4	29.2	30.5	28	27.8
24	26.6	26	25.7	26.5	26.5	25.8	27.3	24.7	25.5	25.7	26.5	29	26.7	26
24.6	27.4	27.5	26.9	26.8	26.8	27.8	29.1	26.4	26.8	25.9	28	29.4	27.1	26.6
22.1	24	24.4	25.5	23.5	23.4	24.3	25	22.9	23.5	21.5	24.7	25.5	23.3	23.6
25.2	27.4	27.2	26.5	27.8	26	26.9	27.7	26.2	26.3	25.1	27.4	28.9	26.2	26.3
24.4	27.1	26.8	27.4	27	26.8	27.9	29	26.6	26.4	26.3	28.2	29.8	27.5	26.9
24.9	27	26.6	27.2	27.2	26.3	27.9	28.2	25.8	26.3	25.9	27.4	28.4	26.8	26.8
23.4	26.5	26.9	27	25.5	25.9	26	27.8	25.9	25.6	24.5	26.9	27.3	26.9	25.7
24.5	25.2	24.9	26	26.2	24.6	25.6	26.6	24.4	25.1	24.7	26.1	27.9	25.5	25.4
23.8	26.7	27.3	27.1	26.6	27.3	27.2	27.4	27	26.6	25.5	27.8	27.8	28.2	26
26.5	27	27.2	28.1	26.5	26.4	27	28.2	26	26.3	25.5	27.6	28.8	27.9	27.8
22	24	24.8	25.3	23.9	24	24.7	25.6	23.2	23.2	22.9	24.8	25.1	24.7	23.5
26.5	28.7	28.6	29	29	28.2	29.2	31.1	28	28	27.3	28.8	30	28.2	29.1
24.2	25.7	27.2	26.8	25.9	26.6	27.3	28.1	26.8	25.4	24	26.2	27	27.2	25
25.4	28	28.6	28.3	28.3	27.6	28	29.4	26.3	26.8	26.2	28.1	30.1	29	26.9
23.1	25.7	27.1	27	25.7	26.6	27.6	26.5	25	24.8	23.9	26	26.6	25.6	24.7
24.9	27.4	28	26.9	27.2	26.4	27.2	28.6	25.9	26	25.8	27.3	29.1	27.8	26.5
25.1	27.5	27	27.7	27.9	26.9	27.7	29.1	26.8	26.5	26.7	27.8	29.8	27.4	26.6
24.3	27	26.6	27	25.9	26.1	26.8	28.1	25.8	26.4	25.8	28	28.2	26.8	26.2
25	28	25.5	25.6	27.3	26.6	27.8	29.6	25.6	26.4	27.5	28.8	30.5	27.1	25.6
25.7	27.7	25.4	25.8	27.6	26.7	27.5	28.2	26.4	26.3	27.3	28.9	30.6	27.9	25.8
25.4	27.6	25.2	25.1	27.2	26.1	27.7	28.2	25.6	27.1	27.3	28.4	30	27.3	26
25.2	27.3	25.5	24.8	26.8	26	27	27.9	25.3	25.8	26.3	27.7	29.7	27.2	25.4
25.6	27.6	25.9	25.5	27.4	26.9	27.6	28.6	26.2	26.5	28.2	28.1	30.3	27.8	25.9
27.3	29.2	27.2	26.5	28.9	28.1	28.2	29.7	27.3	27.5	29	29.2	31.6	28.7	27.1
29.2	31.5	29.5	27.6	30.4	29.6	30.5	32	28.6	29.6	31.3	31.7	33.2	29.9	29.4
27.3	29.2	26.8	26.1	28.5	27.4	28.6	29.8	27.1	27.8	28.6	29.2	30.9	27.5	26.6
28.2	30.1	28.1	26.9	29.7	28.5	29	30.6	28.3	28	30	30.2	31.7	28.7	28.8
27.7	29.2	26.9	26.4	28.7	28.7	29.8	30.7	27.3	27.8	28.6	29.8	31.1	28.3	27.2
28.5	29.5	27.9	27.1	29.5	28.7	28.7	30.8	28	28	28.9	29.6	32.2	28.9	27.8

30	31	29.5	28.7	30.9	30.1	30.4	32.6	29.3	30.2	31	31.1	32.6	29.1	29.8
28.1	29.4	27.2	26.5	28.9	28	29.1	30.3	27.2	28	28.8	29.4	31.2	28.1	27.2
28.2	29.8	27.7	27.8	30.3	29.9	29.1	30.5	27.4	28.3	28.9	29.6	32	28.9	27.4
28.4	29.5	27.7	26.8	29.5	28.4	28.8	30.1	27.6	27.5	27.9	28.7	31.1	27.8	26.5
27.7	30.3	27.9	27.2	29.7	28	29.2	30.3	27.7	28.3	29.2	30.1	31.3	27.7	27.5
28.4	30.6	28.4	26.5	29.5	28.6	29.5	31.2	27.4	28.5	28.1	29.2	31.2	28.5	27.8
28.5	30.4	28	27.1	29.5	29.3	29.6	30.6	27.3	28.6	29.2	29.7	31.5	29.1	28.4
29.1	31.1	28.7	27.6	29.7	29.5	30.5	31.4	28.5	28.8	29.6	29.8	32.7	29.4	28.8
28.5	31	29.2	28.2	30.5	29.5	30.5	31.6	29	29.4	30.1	30.6	31.7	29.3	29
27.5	30	28.3	27.9	30.5	29.2	30.4	32	28	28.8	29.5	30.4	32.3	29	28.1
29.2	30.5	28.5	27.4	29.6	28.5	28.6	30	28.2	29	29.2	28.7	31.6	28.4	27.2
28.7	30.7	28.9	27.9	30.1	29.8	30.2	31.8	28.2	29.3	29.9	30.1	32.4	29.5	28.2
28.5	31.2	28	27	29.8	29	29.1	30.9	27.6	28.2	29.2	29.5	32.1	28.8	27.4
29.2	30.7	28.4	27.9	29.9	29.1	29.4	31.3	27.3	28.2	28.8	29.3	32	28.8	28
29.5	32.8	28.9	27.9	30.5	29.1	29.9	31.9	27.8	28.9	29.6	29.4	32.4	30	29
32.7	36	37.8	33.7	38.8	35.4	33.8	35	35.5	34.8	35.7	32.7	34.4	34.1	33.1
31.5	33.7	34.4	31.3	35.9	32.5	30.5	33.5	32.5	33.5	32.9	32.7	31.5	31.3	31
32.7	34.8	36	33.8	38.5	33.8	33.2	35.8	34.4	35.2	35.2	33.1	32.9	32.9	33.3
31.8	33.3	35.2	31.1	35.9	33.4	31.6	33.7	33.6	34.3	33.5	32.3	31.5	31.4	31
33.4	36	38	33.7	39.3	35.3	33.9	34.7	35.5	34.1	36.3	31.4	34	33.6	32.8
33.7	37.4	39.2	35	39.9	37.3	35.4	37.1	36.6	36.1	37.3	34.1	34.4	34.9	33.6
32.1	34.6	37.2	31.8	37.7	33.5	33.3	32.7	33.5	32.9	35.5	31.6	34	30.9	32.5
33	37	38.2	33.7	39.4	36.4	33.8	36.4	36.1	34.5	36	32.7	32.8	32.7	32
34.5	35.8	37.1	33.5	38.6	35.1	35.1	35.2	35.4	35.5	37.1	33.5	34.4	32.3	34.2
33.6	35.1	38.2	33.2	38.6	34.6	33	34.7	35.2	34	35.3	31.4	33.4	31.9	31.8
33.8	36	37.8	33.2	40	35.4	34.4	35.6	35.4	34.7	37.3	32	34.7	32.5	33.9
32.8	36.7	38.6	34.9	39.7	36.4	34.4	35.7	36.2	34	36.4	33.1	35.1	34	33.9
32.3	35.3	37.4	33	39.9	35.4	33.7	33.8	34.9	34.4	36.4	32.5	34.6	32.8	33.2
33.6	35.1	38.1	33.8	38.8	34.4	33.2	35	35.5	35.3	36.3	32.6	34.1	33.4	32.8
33.3	35	36.2	32.7	37.8	34.2	32.6	34	34.4	33.6	36.6	32.3	34.6	32.1	33.2
32.7	36	37.6	34.3	39.9	35.5	33	34.6	35.1	34.2	36.4	32	34.4	31.8	32.8
32.2	36.5	38.2	34.2	39.6	35.6	34.1	34.2	36.4	34.4	36.7	32.1	33.8	32.6	33.1
32.9	34.9	37.7	32.4	39.2	33.2	33	33.7	33.8	34.6	36.5	33.1	34.2	31.3	33.7
32.2	35.1	37.3	32.3	38.5	33.7	32.4	33.5	34	33.6	35.8	31.3	33.3	31	33
34.2	35.9	38.6	34	39.5	35.1	33.9	35.6	35.9	35.6	37.2	32.2	34	32.5	33.9
33.7	36.2	37.6	34.3	38.8	35.1	34.2	36	35.1	35.6	36	34	33.8	32.9	34.2
33.8	36	37.7	33.7	39.1	35.8	34.4	35.2	36.1	34.9	37.2	33.4	33.9	32.8	33.9
31.6	34.9	35.9	32.2	38.3	34.1	32.6	33.4	33.3	32.8	35.9	31.4	33.7	31	32.2
32.9	35.5	37.5	33.1	39.6	33.8	33.7	34.5	34.6	34.9	36.5	32.7	34.2	31.4	34.5
33.7	36.5	38	34	39.3	35.6	34.5	35.6	35.8	34.8	35.6	32.4	35.2	34.1	34
33.9	35.8	37.6	34	39.4	34.7	33.8	36.2	35	35.3	35.5	32.7	33	32.1	33
34.1	36.2	38.3	34.1	38.5	35.4	34.4	36	35.9	35	37.5	33.2	34	32.2	33.2
33.1	36	38.5	35.3	40.5	36.7	34.3	36.2	36.3	35.3	37.1	33	35	34.1	33.1
34	35.5	37	33.2	38.4	34.8	33.8	35	34.5	33.1	35.1	32.1	34.1	32	33.1
34	37.1	38.7	34.2	39.8	35.2	34.6	36.8	36.6	36.8	37.2	33.8	35.6	33.4	33.8
33.8	35.6	37.8	34.1	38.8	34.8	34.2	35.5	35.4	34.6	36	32.5	33.9	32.4	33.3
32	35.8	36.3	32.1	38.4	34	32.8	33.2	34.7	33.8	36.1	31.9	33.5	30.4	32.1

33.2	36.5	39	34.4	40.5	36.3	34.8	35	36	34.8	36.8	32.7	34.3	31.7	33.9
34.5	36.8	39.2	34.4	40.3	36	35.4	36.4	35.8	36.1	36.4	33.3	34.1	34.2	33.3
33.6	37.3	39.3	34.5	40	37.2	34.8	36.8	37	36.9	37	33.7	35.6	34.4	33.1
34.1	36.1	38.5	34.1	39.2	35.2	33.9	35.3	36.1	35.5	37.1	32.9	33.7	33.1	34.3
34.4	36.9	38.5	34	40	35.8	34	35.7	36.2	35.6	37.3	32.8	35.4	32.8	33.3
32.6	35.1	36.3	33.4	39.2	34.7	33.4	34.4	34.3	34.1	36	32.5	33.9	31.7	33.3
32.5	35	36	31.6	37.9	33.6	33.2	33.3	35.3	34.5	36.8	32.2	33.2	31.3	33.8
32.2	35	36.6	32.4	38.2	34.3	33.6	35	34.5	34	34.7	33.7	33.9	31.5	33
32.2	34.1	36.8	32.8	39.9	33.8	34.5	34.9	35.6	34.6	35	32.5	31.9	30.7	31.2
32.3	35.4	36.3	32.7	38.1	35.1	33.6	34.7	34.4	33.5	35.3	31.7	33	30.7	32.3
33.8	35.9	38.3	33.6	38.9	35.7	34.3	35.7	34.9	35.5	36.2	32.5	33.5	32.7	33.6
32.3	35.5	37.9	32.9	39.2	34.6	33.3	34.8	34.8	33.8	35.7	31.7	32.9	31.1	33.3
25.8	25.5	26.3	27.3	27.1	25.4	24.7	22.6	28.2	30.2	24.3	24.5	27.6	26.4	28.5
30.7	30.6	30.5	30.4	31.3	31.3	29.8	26.5	33.6	34.4	29.9	29.7	33.8	30	34.3
26.9	27.5	27.1	28.9	27.9	26.7	25.7	23.1	27.8	30.3	24.7	24.8	28.8	25	29.3
29.4	28.1	28.2	29.9	28.3	27.2	26.5	25.8	29	32.7	26.1	27.3	29.9	29.7	30.8
25.7	25.6	24.8	26.8	26.1	26.1	24.1	23.4	26	27.5	24.5	23.1	27.8	26.7	27.7
27.9	27.2	27.2	27.7	27.7	27.2	24.9	23.9	28.1	29.9	26	23.9	29.7	29.1	29.1
29.5	29.7	29.7	31.1	29.5	28.9	27	24.9	29.8	32.7	27.7	27	30.5	26.5	32.1
27.9	28.4	29.6	29.8	29	28	28.2	27.6	31.4	34.4	26.8	27.7	30	27.1	32.1
24.5	24.8	25.7	26	25	24.2	25.1	22.8	26.5	29.4	22.6	23.5	26.2	23.4	28
29.2	28.3	27.8	29.5	28.6	28	27.2	26.6	29.3	31.4	27.3	27.3	31.6	30.5	30.3
26.4	26	27.5	28.3	27.1	26.3	26.5	24.1	28.4	31.7	24.2	25.5	28.6	25.7	30.6
29.4	29.8	29.5	29.5	29.8	29.7	28.2	25.4	30.8	31.9	27.7	26.6	31.4	27.5	31.3
31.2	31.6	30.7	31.8	30.4	31.6	28.4	29.2	33.2	34.4	29.8	30.4	34.7	32.2	32.7
29.7	29	28.6	30.4	28.7	28.1	27.5	26.9	30.2	32.7	26.4	27.1	30.1	30.3	31.7
19.5	19.2	19	19	19.5	20.6	17.2	12.9	20	22.1	16.9	16.8	21.1	18	20.6
28.1	26.8	27	29.3	27.4	26.8	25.7	23.2	28.6	30.2	26.4	25.2	28.9	24.9	29.4
28.2	27.9	28.5	29.6	28.5	27.9	26.3	26.3	28.9	32.3	25.9	26.3	29.1	29.5	29.9
24.3	23.8	25	26.2	24.7	24.1	22.8	20.8	25.6	29.2	22.4	23	25.6	22.6	26.8
26.4	26.2	27.3	28	27.2	26.7	25.7	23.5	27.5	30.8	24.9	25.6	28.2	25	29.5
28.7	28.9	27.4	29.1	27.9	28.5	26.2	25	30	31.3	26.6	25.2	30.8	28.6	29.8
27.7	27	27.7	28.9	27.8	26.6	26.7	24.7	28.4	31.5	25.2	25.6	27.3	26.8	29.1
30.8	30.1	30.2	30.5	30.8	30.4	28.6	27.5	31.5	32.7	28.7	27	32.5	30.2	31.7
26.8	26.5	29.2	28.1	27.6	27.3	26.8	24.8	28.4	30.7	24.5	24.5	28.8	25.3	29.5
29.6	28.8	29.9	31.2	30.1	28.6	29.3	27.6	31	34.1	27.2	28	30.5	29.7	32.6
28.9	28.5	28.3	28.8	29.4	29.4	26.9	23.8	29.1	30.9	26.7	25.6	30.4	26.2	30
31	30.4	31.9	32.5	31	29.4	30.8	28.4	31.5	34.9	28.6	30.4	32.3	31.1	34
24	22.7	24.6	25.2	24.2	24.3	23.3	20.6	25.4	28.5	21.6	22.4	25.3	22.9	26.2
26.2	25.9	26.1	27.8	25.9	25.3	24.1	22.5	26.8	29.5	24.5	24.2	27.8	24	27.9
30.4	28.7	29.8	28.9	28.8	26.4	27	27	30.9	30.7	25.2	23.5	30.8	28.7	29.6
29.7	29.9	30	31.2	30.1	29.6	28.5	25.7	31.4	33.7	28.5	27.7	31.4	27	32.1
32	31.3	31.6	32.8	31.2	30.3	30.7	28.4	32.4	35.4	28.1	29.8	31.9	30.8	34
22.3	22.3	22.6	23.5	23.2	22.2	21.7	19.9	25	27.1	20.8	22.4	24.2	21.6	26.2
30.1	29.2	29.3	32.1	30.2	29.2	27.6	25.6	30.9	33	26.7	26.1	31.4	29	32.1
25.8	25.6	26.9	27.8	26.4	25.6	25	23.6	27.3	31.1	24	25.2	27.7	24.6	28.8
29.2	29	30	31.2	30	28.4	28.1	26.9	31	33.2	26.4	27.4	29.8	28.6	31.5

16.5	17.2	16.9	17.5	18.1	17.6	16.6	17.9	19	19.2	18.8	17.6	18.6	17.4	18.3
20.1	19	19.7	20.9	19	19.1	18.6	19.6	19.9	20.5	18.5	18.8	20.3	19.3	21.1
19.4	18.8	19	19.8	18.6	18.6	17.3	19.4	18.7	19.1	17.4	18.1	18.6	18.5	19.8
18.8	17.9	18.5	19.2	17.9	18.5	18.2	19.2	17.8	18.5	17.9	18	18.8	19.3	19.1
17.9	17.2	17.3	18.8	17.2	16.4	16.6	18.4	16.5	17.5	16.1	15.7	17.3	18	17.6
22	25.1	23.5	24	25	24	24.7	25.3	23.9	23.9	24.6	26	27.7	24.9	25
21.9	23.8	22.9	24.4	25.2	23.4	24.2	25.1	22.5	24	23.8	25.3	27.2	24.5	24.3
25.9	28.4	26.9	26.9	29.1	28.5	28.1	29.2	27.3	27.3	27.7	30.2	32.3	27.9	28.2
27.2	29.8	27.4	28	29.1	28.9	29.6	30.1	27.5	28.8	29	30.2	31.5	28.8	27.5
26.9	29	26.8	26.8	30.2	28.6	28.2	30.3	27.1	27.6	27.9	29.3	30.8	27	28.2
26.7	29.5	27.8	28	31.2	28	29.4	31.2	27.9	30.1	30	31.5	32.1	28.3	29.6
26.5	28.9	26.7	26.9	29.6	27.7	29	30.1	27.3	28.5	28.3	30.7	32.2	28	28.3
27.7	29.9	27.8	28.1	30.7	29.5	29.8	30	28.6	29.8	30	31.5	31.6	28.5	29.4
23.4	26	24.9	26	26.5	25.2	26.3	27.1	24.4	25.5	24.8	27.5	28.6	25.5	25.4
27.4	30.8	27.5	27.6	30.8	29	29.6	31.2	28.1	29.2	31	31	32.6	28.2	29.2
25.9	28.4	26.5	26.4	29.3	28.8	28	29.7	27.2	28.3	28.5	30.8	31.3	27	28.4
27.5	30.7	28.2	27.6	30.9	29.1	29.6	32	27.9	29	30.8	31.7	33.3	28.2	29.7
25.8	28.2	27	27.6	28.9	27.1	27.6	29.2	26.6	27.3	26.9	29.2	31.4	27.5	27.3
27.7	31.5	28.8	28.7	31.3	29.5	30.3	31.7	28.6	29.1	31.4	32	32.6	29.4	30.1
23.9	27.3	26.1	26.4	27.3	26.9	27.4	28	25.4	26.2	26.8	28.5	29.5	27.7	26.1
25.8	28.3	26.5	26.3	27.4	27	27.4	28.9	26.6	26.7	27.4	28.7	30.2	28	26.9
29.8	32.3	29.9	28.9	31.2	30.2	30.6	31.7	28.5	30.8	30	31.6	33.2	29.8	31.3
26.1	28.7	27	26.4	28	27.7	28.3	29.2	26.6	27.3	26.3	29.2	30.8	28.5	26.9
25.2	28.5	25.9	26.1	28.1	26.9	27.5	28.8	25.8	26.5	26.8	28.9	30.3	27	26.5
24.1	26.4	26.3	24.7	26.3	26.6	26.2	27.2	25.4	25.3	25.9	26.9	28.3	26.3	25.6
28.1	31.8	29.3	28.6	30.4	29.6	30.8	30.6	29	29.5	30.4	32.2	33	30.2	29.1
26.6	30.2	27.3	27.4	29	28.6	29.4	29.9	27.2	28.7	29.3	30	31	28.1	28
25.1	28	26.2	26.9	28	27.6	28	28.6	26.5	26.9	27	28.9	29.5	28.2	26.6
27.5	29.8	27.9	28.2	30.2	28.6	29.2	30.8	27.3	28.9	27.7	29.9	31.3	27.7	28.5
26.1	27.3	25.7	27.2	27.5	27.9	27.7	28.3	26.4	26.9	25.5	27.2	27.1	26.6	26.6
25.1	26.4	24.2	26	27	26	26.9	26.3	24.3	25.2	23.9	25.8	25.7	24.6	25.3
25.5	28.4	26.6	28.2	28.8	30	27.7	29.8	25.8	27.5	25.5	26.6	28	26.7	26.6
31.6	34.4	33	31.1	35	33	32	34.6	31	32.2	37	34.5	33.3	32	35.3
31.8	33.3	32.9	31.7	33.1	33.3	30.7	33.1	30.8	31.9	34.9	33.4	32.5	31.5	33.2
31	34.3	33.2	29.9	33.5	32.3	31.6	34.1	29.9	31.1	36	34.2	32.4	30.4	33.2
29.5	32.7	30.9	27.9	33.3	31	30.2	32.5	27.9	28.8	34.4	31.9	30.3	28.9	31.1
29.8	33.7	31.9	29.2	33.3	31.7	30.4	33.1	29	30.5	34.6	32.3	31.1	30.3	32.4
30.6	33.9	32.4	29.4	34.5	32.2	31.3	33.9	28.7	29.3	34.8	32.7	30.8	30	32.3
32.3	35.2	31.3	31.7	34.7	32.6	31.4	33.7	30.8	31.5	35.9	34.1	32.5	31.6	34.3
32.9	35	33.2	31.7	34.7	33.3	31.8	34.1	31.4	32.5	36.4	33.9	33.3	32	34.2
31.1	34.3	33.2	30	34.8	33	31.3	34.1	29.5	30.5	34.6	33.2	31.6	30.8	32.8
31.2	34.5	32.8	29.8	35.2	31.6	31.3	34.6	29.7	30.7	35	33.4	32.6	30.2	32.8
30.5	32	30.5	30.2	31.2	30.8	30.8	31.2	29.2	31.6	32	30.8	28.7	29.8	30.1
31.9	34.9	32.4	31	34.2	32.9	32.1	34	31.5	32.3	35.4	33.1	32.6	32.4	33.9
31.3	34.7	33	29.8	35	33.1	31.4	34.1	30.5	31	35.7	33.7	32.1	31.2	32.9
31	34.6	31.4	30.4	33.5	31.7	30.6	33.4	29.9	31.1	35.6	32.6	31.2	30.8	33.3
30.3	34.4	31.7	30.4	33.7	31.7	30.9	32.9	30	30.9	35.9	32.7	31.7	30.1	33.6

31.3	33.9	32.1	30.2	33.9	31.8	31.5	33.9	29.7	31.2	35.7	33.6	33.4	30.4	32.8
31.6	35.3	33.8	31.2	35.3	33.2	32.3	35	30.6	32	36.2	33.9	33.7	31.3	33.9
33	34.2	33.3	32.7	34	33.8	32.1	34.1	31.9	32.1	35.2	33.4	32.9	33.2	33.6
32	34.2	32.4	31.9	33.3	32.8	31.8	33.6	31.2	31.9	35.6	33.9	32.6	32.8	34.1
31.4	35.4	32.6	31	34.6	32.5	32	34.4	30.9	31.8	36.3	33.9	32.3	31.6	33.3
32.9	35.1	34.1	31.6	35.5	34	32.5	34.8	31.6	32.4	37.2	35	34.2	32.2	34.8
31.4	32.8	31.5	31.2	32.4	32.5	30.8	32.5	30.3	31.2	34.7	33.3	31.4	31.9	33
32.7	36.1	34.3	31.7	36.6	33.9	33.5	35.6	31.4	32.2	36.5	35	33.7	31.4	34.5
32.4	34.9	31.9	29.6	34.9	33.2	31.1	34	30.3	30.6	36.2	34.3	32.6	31.6	32.8
32.8	36	34.6	31.4	36.8	34.2	33.1	36.1	31.6	32.6	37.1	35.3	34.1	32.6	34.9
32.8	37.5	34.3	33	34.5	33.5	32.6	34.3	31.9	32.5	37.4	35.2	34.1	32.3	35.5
31.7	35.1	32.1	31	33.5	32.6	31.6	34	31	31	34.7	33.3	32.2	30.4	33.1
32.4	35.4	32.9	32.8	35	33.4	33.5	34.5	32.1	32.6	37.4	35	33.4	32.7	35.3
31.7	34.9	32.8	30.4	33.8	32.2	31.1	34.5	30	30.9	36	34.3	32.2	31.6	32.9
28.6	30	27.5	26.9	31.6	28.5	27.9	30.9	29	30	27.7	29.8	31	31	27.6
33.2	30.1	28.5	27.8	33.6	29.6	29.5	31.2	29.7	32	29.8	32	32.1	32.3	27.9
32.3	32.3	31.4	28.7	34.7	30.2	31.2	32.8	31.1	34.6	30.7	32.7	31.9	34.3	30.4
31.1	30.3	27.3	27.1	31.7	28.9	28.6	29.2	27.3	28.1	28.3	29.7	31.2	30	27.2
31.8	32.1	31.1	30	34.2	30.7	31.7	33.1	31.3	33	30.8	32.6	33.4	33.3	30
33.9	31.2	29.3	28.9	34.2	30.2	30	31.7	30.5	31.6	30.9	31.4	31.3	32.4	28.9
32.2	30.3	28.4	28.3	30.6	28	29.6	30.5	28.1	28	28.9	29.2	30.8	30.2	27.3
34.6	31.9	30.9	29.6	35.2	32.2	31.1	32.9	31.4	32.4	32.2	33.4	33	33.2	29.3
34	33.2	31.1	30.1	36.5	32.8	32.3	34.6	32.6	35.2	31.8	33.1	34.7	34.3	31.3
34.2	32.9	30.3	28.4	35	30.8	32	33.1	32.2	33.7	30.9	33.5	33.5	34.1	30
33	31.3	29.3	28.8	31.3	30.9	30	30.9	29.2	29.6	30.5	31.5	32.7	30.8	27.8
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32.9	31.8	29.6	29.1	33.4	29.6	30.4	32	29.4	29.7	30.7	30.6	32.2	31.9	29.7
34.1	32.3	30.7	30.2	35	31	31.7	34	31.9	33.5	31.4	34.9	34.3	34.6	30.4
32.8	30.7	28.6	28.8	30.6	29.3	30.9	32	29.8	29.9	29.3	30.9	32.5	31.7	27.6
32	29	27.4	27.2	28.9	28	29	31	28.6	28.8	28.8	30.2	31.6	30.8	27
32.3	29.8	27.2	26.5	29.2	29.8	28.8	30.4	28.4	28.2	28.6	30	31.7	30.4	26.7
33.7	32.6	32	30.9	36.7	32.1	32	33.4	31.6	32.9	32.2	33.2	32.9	33.8	29.3
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30.9	33.8	32.4	29.9	34.7	32.3	32.4	33.5	30.6	31.7	34	32.5	32.3	30.9	32.2
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30.2	32.4	31.7	28.9	33.1	31.2	31.7	32.4	29.5	31.1	33.4	32.4	31.7	30.3	31.3
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32.1	34	33.3	31.1	35.8	32.6	33.3	34.5	31.8	32.6	35	32.2	31.9	30.7	32.8
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28.7	32.7	33.4	33.1	35.5	32.5	32.7	33.9	30.5	31.9	32	31.9	30.2	33	28.8
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32.6	36.1	36.4	37.8	38.8	36.1	37	36.3	34.6	34.9	34.3	35.7	34.5	35.5	34
31.9	35.4	37.1	33.7	38	34	37	34.1	37.2	35	36.9	34.3	36.1	36.5	34.7
33	37.2	37.5	37.5	39.6	37.2	37.9	35.2	36.2	36.6	37.3	35.8	35.8	37.7	33.3
29.9	33.6	35.2	32.2	36.4	33.5	32.6	33.7	31.8	32.7	33.7	31.6	30.2	32.2	30
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31	33.4	33.9	32	34.8	32.7	33.8	31.3	32.3	32.4	33.6	32.4	32.8	31.8	32.4
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31	35.4	37.6	32.8	37.5	34.3	33.1	33	34.9	33.7	34.5	32.4	33.3	33.4	32.6
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33.7	38.7	39.3	38.2	40.9	36.8	39.2	37.1	39.1	37.4	39.8	36.6	37.7	39.6	36.9
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29.1	33.3	34.3	32.6	35.6	32.8	32.4	35.1	33	34.2	33.8	32.4	30.8	32.1	30.9
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33	31.8	33.4	32.8	33.1	32.7	30.5	30.4	31.3	32.1	31.7	30.9	34.3	34.5	32.9
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22.8	22.4	24.8	24.5	23.6	22.6	22.1	20	22.4	23.1	21.5	20.8	23.8	24.6	22.6
32.4	31	31.7	31.6	31.9	31.5	29.8	29.7	30.2	31.5	29.5	29.4	33.7	34.2	32.4
33	31.7	32.2	32.5	31.7	31.5	29.5	29.6	29.9	31.3	29.8	29.8	33.5	34.6	31.8
34.3	32.5	34	33.3	33.7	33.4	32	31.6	32.5	32.6	31.1	31.9	35.9	36	34
27.7	27	28.6	28.4	29.1	28.5	26.5	26.5	26.8	27.9	25.9	26.2	28.5	30	28.9
32	32.6	33	33.9	32.4	34.3	31.6	31.7	31.6	33.7	31	30.9	34.3	34	32.8
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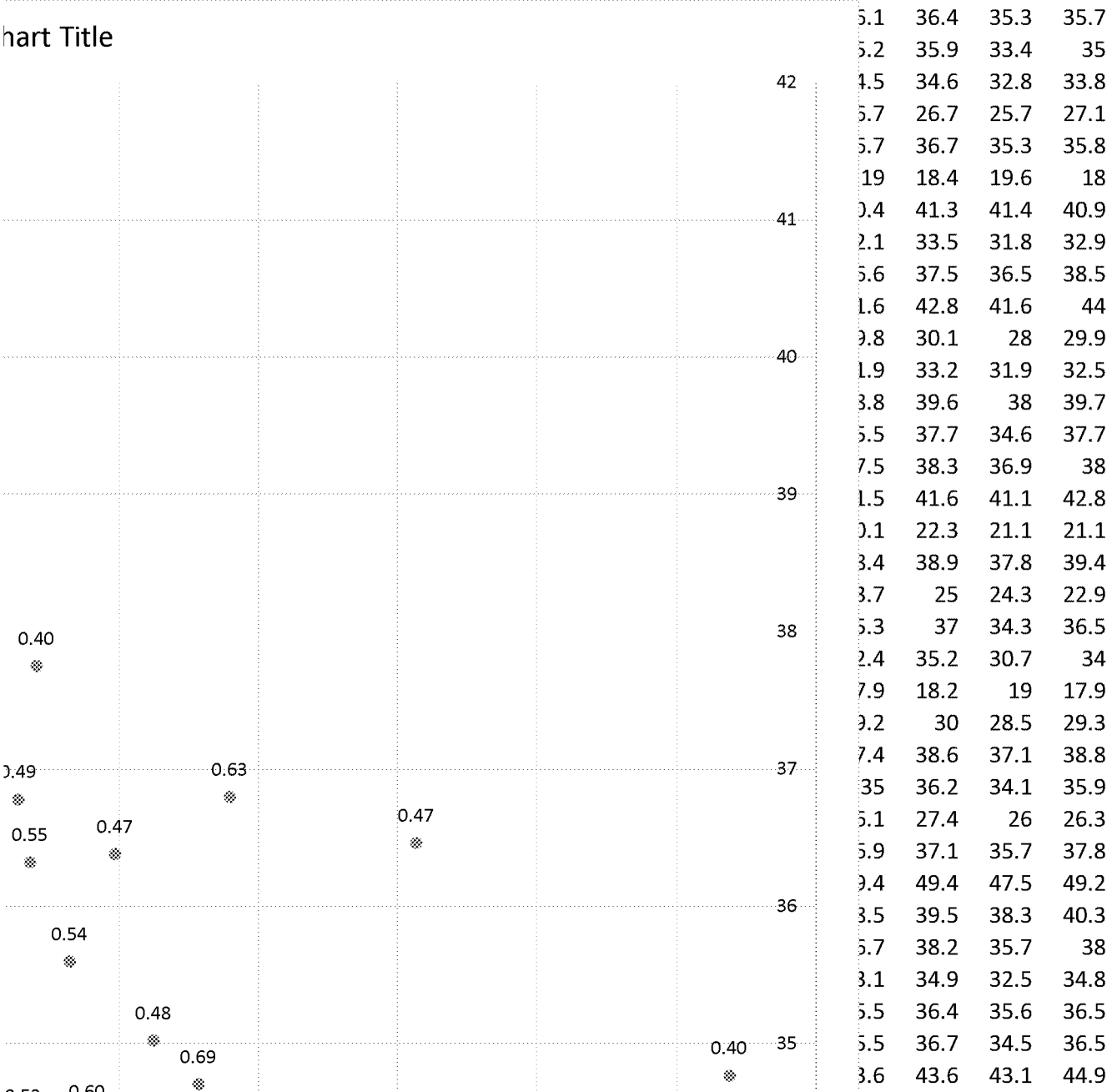
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27.1	30.3	27.7	27.2	29.8	28.2	28.4	30.4	27	28	31.4	31	31.8	27.6	28.6
32.3	35.4	31.5	31	33.9	32.1	32.4	34.1	30.6	31.9	33.6	34.9	32.5	32.1	32.6
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28.4	32	27.6	27.2	30.9	27.6	28.3	31.1	27.2	28.6	31.3	31.3	31.3	28.1	29.8
24.1	26.4	24.8	23.4	26.9	24.7	25.3	27	23.6	24.2	26.4	27.4	27	24.7	25.7
31	35.1	31	29.7	33.7	31.6	31.7	33.3	30.2	31.9	33.5	35.1	33.7	31	32.5
30.1	32.1	29.3	29	32.5	30.7	30.1	32.3	28.2	30.1	32.5	32.5	31.5	30	31.6
28.4	32.7	28.6	27.6	30.7	28.6	28.8	30.7	28.1	30	31.1	32	31.8	28.6	29.6
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30.2	33.2	29.4	29.4	32.2	30.2	30.8	33.2	29.2	30.7	31.7	33.3	31.9	29.4	30.9
29.5	33	29.6	28.4	31.5	29.3	30	31.2	27.9	28.6	32.5	32	31.1	28.3	30.1
32.5	34.9	31.5	30.5	33.2	31.7	31.9	33.7	30	31	34.4	34	33.1	31.5	32.9
29.1	32.5	29.5	28.2	31.5	29.6	30	31.6	28.3	28.8	31.6	32	31.9	29.5	30
27.1	29.8	27.7	26.4	29.6	27.1	28.4	28.8	25.9	27.1	29.6	30.4	30.2	28.2	29.4
30.6	32	30.7	27.5	33	31	30.5	31.6	29.1	28.2	32.3	31.7	31.8	29.5	30.5
25.7	27.7	27.8	28.3	27.5	26.6	27	27.1	26.4	26.3	25.1	26.5	27	26.7	26.5
21.9	23.9	23.9	24.4	24.6	23.6	24.3	24.5	23.6	24.1	21.7	24.4	25.4	24.6	24.1
25.8	27.1	27.6	28.5	27.6	27	28.2	29.3	26.5	27.6	25.5	28	29	27	27.9
25.3	26.9	26.6	27.4	27.3	26.7	27.3	28.1	26.3	26.5	24.7	26.7	27.6	26.3	26.8
24.4	26.2	26.5	27.3	26.4	26	26.4	26.9	25.7	26.8	24.4	27.1	27.4	26.8	26.9
24.6	26	26.6	27.7	25.7	26.2	26.5	26.7	26	25.6	24.4	26.7	26.9	26.6	26
26.4	27.4	27.5	29.2	27.4	27.5	27.9	28.3	27	27.2	25.9	28.3	28.9	28.7	27.5
25	27.1	27.1	27.7	26.6	27.1	27	26.8	26.5	27.1	24.9	28.1	28.3	27.3	27
26.2	27.8	28	29.9	27.7	25.3	26.1	23.8	28.2	33.6	25.3	27.3	28.8	28.9	27.8
25.7	25.3	26.7	27.4	26.4	25.7	25.9	24	27.8	31	24.3	26.1	27.1	27.2	28.7
28.3	28.7	29.4	30.4	29.8	28.6	28.7	26.9	31	34.1	27.4	28.4	30.9	29.1	32.4
21.6	21	21.9	22.7	22.2	21.1	22.3	17.7	24.3	28.3	19.3	21.4	23.6	21.6	24.6
22.7	23.1	24.4	25.3	24.1	23	22.5	21.2	24.3	28.5	22	23	25	22.3	26.5
27.7	28.6	29	30.8	28.1	26.6	26.9	24.5	30	33.9	26	27.4	29.9	30.9	29.9
29.7	28.5	29.9	31.2	30.3	28.7	29.5	27.7	31.5	33.7	27	28	31.1	30.1	32.9
31.2	30.8	30.9	32	32	30.6	31.2	29.3	32.8	36	28.9	30.4	32	31	33.6
27.9	28.5	29.7	29.7	28.8	27.9	27	25.6	29.5	32.9	26.2	27.5	29.6	28.3	31.6
27.8	26.9	27.3	29.2	28.7	26.5	27	24.9	29	32.9	24.9	26	28.4	28.6	29.2
27	26.6	27.3	28.4	26.6	26.3	26	24.6	28.9	32.6	25.1	26.8	27.6	28.4	29.7
26.6	26.9	28.6	29.6	28.6	26.4	25.8	23.2	28.7	33.7	24.8	27.1	28	27.5	29.9
24.6	24.2	25.5	26.9	25.8	25.1	24.4	23.4	26.1	29.6	23.4	24.5	26.2	24	27.8
27.8	28.6	28.9	30.1	28.8	27.7	28.1	25.7	30.2	33.4	25.7	27.4	29.1	26.8	31.6
20.2	19.8	21.7	21.9	20.4	21.3	20	20.4	22	25.6	20.3	20.8	21.2	20	23.3
26	26.5	27.3	30	27.9	25.9	26.1	24.1	28.8	32.5	24.2	25.8	27.9	27.6	27.8
22.5	22.4	23.8	25	23	22.1	22.3	22.2	24.4	28.2	21.2	23.2	24.4	22.2	25.6
20.5	19.2	21.9	21.8	20	19.6	19.1	19.5	21.4	24.6	18.2	19.9	20.3	19.6	21.8

27.5	26.5	27.4	28.8	27.4	26.2	25.5	23.4	27.5	32.2	24.4	26.5	27.5	27.6	28.9
27.9	28.2	29.4	30.4	28.6	27	26.4	25.2	29.5	33.3	25.7	27.5	28.6	28.4	29.2
27.7	28	28.6	30.1	28.5	27.1	26.9	24.6	29.4	33.2	25.4	26.1	28.8	28.7	30.7
29.4	28.4	29.4	30.7	30.2	28.9	29.3	26.9	30.9	33.7	26.2	27.7	29.8	29.6	31.9
24.9	24.6	26.3	28.1	27	25.3	26.8	24	26.8	30.3	22.6	24.5	26.2	25	29
27.2	27.5	29	29.8	28.8	27.7	28	25.4	29	33.4	25.2	27.4	27.9	27.1	29.9
23.8	22.8	24.1	24.4	22.8	23.2	22.3	22.6	24.2	27.2	21.9	23	24.3	22.7	25.6
22.1	21.1	23.5	24.4	21.9	21.3	21.2	21	23.2	26.6	20.5	22.6	22.2	21.6	24.4
21.7	22.1	23	23.7	22.7	21.6	21.5	20.4	23.7	27.4	20.4	21.4	22.6	21.1	25
19.2	18.7	20.5	21.3	19.2	18	18.4	17.5	21.5	25.8	17.1	19.3	20.9	18.7	23.2
23.1	23	24	25.6	23.9	22.5	22.2	20.9	24.6	28.5	20.8	23	24.4	21.7	25.8
23.2	23.3	24.9	25.3	23.4	23.1	22.8	22.4	25.6	29	21.6	23.3	25.1	22.5	26.7
21.8	21.4	24	24.6	22.3	21.9	22	21.4	24	27.3	20.8	22.5	23.5	22.4	25.7
19.8	19.3	20.4	20.3	18.5	18.9	18.2	19	20.5	21.4	17.4	18.9	19.6	19.6	21.1
26	26.6	27.9	28.7	26.9	26.6	26.3	24.2	28.1	32.3	25.3	25.5	27.5	26.7	29.1
29.3	28.4	30	31	29.9	28.9	28.8	26.6	30.6	34.6	26.9	28.4	30.1	30	31.7
30.5	30.6	32.1	33.4	31.5	30	29.8	28	32	35.1	28.4	29.8	31	31	33.1
29.5	28.4	29.8	30.4	28.5	27.9	27.5	26.4	30.2	33.2	26	27.3	29.1	29.6	32.7
19.9	20.5	23.3	22	20.9	19.7	19.3	20.6	22.7	25.6	18.5	20.3	22.1	19.9	24.6
18	17.1	19.5	19.2	18.1	17.8	17	17.9	19	20.1	15.2	17.1	18.8	17.3	20.4
20.6	20.3	22	22.5	21.2	21	21	21.2	22.1	24.2	19.7	21.5	22.3	22.2	23.5
23.4	22.9	24.6	25.6	24	23.1	23	22.3	25.1	28.4	21	23	24.2	22.6	25.8
23.4	22	23.8	25.4	22.8	22.9	22.8	21.7	24.5	27.6	21.9	23.4	23.4	23.1	25.3
18	17.6	18.1	18.8	18	17.9	17.6	19.3	18.6	18.9	17.3	17.6	18.2	18.7	19.4
21.6	21.5	21.8	22.3	21.5	20.6	20	21.1	22.3	23.9	18.7	20.3	21.7	20	23.4
20.3	19.6	20.4	20.6	21.5	19.5	19	20.6	20.5	21.2	18.1	19.2	20.7	19.8	21.9
29.1	28	24.2	26.5	27	25.6	26.8	28.7	25.8	26.1	26.4	27.7	29.4	27.4	25.5
28	26.3	24.6	25.5	29	25.9	26.1	28.5	26.1	26	25.2	27.3	30.8	27.3	26.8
28.6	28.2	25	26.3	28.3	26.8	27	29.3	26.2	27.4	27.2	28.3	29.7	28.5	26.1
25.9	25.4	22	24.4	24.6	23.9	24.6	26.7	23.7	23.7	24.3	25.2	27.9	26	23.7
28.7	28.6	24.9	26.5	27	26.1	26.8	29.4	26.5	26.6	27	28.1	28.9	27.4	25.5
30.4	29.5	26.2	27.9	29.6	27.3	28	31.1	27.6	28.1	28.9	29.8	31.7	29.8	27.1
29.6	29.8	26.1	27.9	29.4	27.4	28.4	31	27.3	28.4	27.4	28.9	31.8	27.9	27.5
27.6	28.2	24.8	26.1	27.7	25.3	23.7	27.9	25	26.3	26.2	27.5	27.9	24.6	24.9
28.2	26.7	23.7	26.8	26.8	25.3	26.6	28.5	25.4	25.5	25.1	27.4	29.4	27.5	25.1
30.3	29.6	26.4	27.6	29.3	27	27.8	30.3	27.3	28.1	28.9	29.8	31.6	29.5	26.7
30	29.3	26.5	28.2	29.2	28.7	29	29.8	27.5	28.4	27.9	29.4	30.2	28.3	26.9
29.9	29.7	25.8	27.3	29.2	27.1	27.8	29.7	26.6	28.5	27.8	29.9	30.8	28.6	26.8
29.5	27.6	25.1	26.5	28.5	25.8	28	28.5	26.1	26.4	26.6	27.8	31.1	28.7	26.3
28.8	28	24.5	25.7	27.5	25.6	27.3	29.4	25.4	26.2	26.6	27.8	30.1	27.7	25.7
28.9	28.6	25.9	27	28	26.6	27.4	29	26.6	27.1	27.2	28.4	29.6	27.2	25.9
29.9	28.2	25.4	27.1	27.9	26.4	28.1	30.3	26.9	28	27.4	28.8	30.7	30.2	26.7
30.5	29.8	25.8	27.9	28.7	28.1	28.2	30.5	27	27.8	27.8	28.5	30.5	28.9	26
29.5	28.7	24.9	27.2	28.9	26.6	27.8	29.8	26.1	27.6	27.3	29	30.8	29.1	26.7
30	28.9	26	27.2	29	26.8	27.3	29.6	27.4	29.3	27.9	29.5	31.5	29.7	27
30.4	29.7	26.6	28.4	30.3	26.9	28.1	30.8	27.6	28.5	27.5	28.9	30.9	28.2	27
27.1	26.1	23	25	25.4	25.1	24.6	27.4	25.5	24.6	24.8	26.5	26.5	24.8	24.4

29.1	28.2	25.6	26.3	28.6	26.9	27.6	28.5	26.9	28.5	27.6	28.5	30	28.4	26.2
31.8	31.3	27.9	28.7	31.6	29	29.8	31.8	29.2	31.2	30.4	30.2	32.1	30.1	27.8
27.1	30.5	28.1	26.8	29.8	28	29.3	32.3	28.2	28.2	30.4	31.3	32.5	27.9	28.7
26.3	28.3	26.8	25.8	29.1	28	27.6	28.9	25.6	27	29	29	30	27	27.5
30	31.2	30.3	27.6	28.8	31.6	30.9	31.9	29.4	30.2	33.2	32.7	34.2	31.8	31.7
26.5	29.1	27.2	26.6	28.6	28.1	28.7	29.4	26.2	28.1	29.3	30	31.2	28.3	28.4
25.9	28.7	26.3	26.2	28.5	27.8	28.7	29.1	26.8	27.7	29.5	30.5	31.1	28.7	28.2
23	25.9	24.3	22.5	25.6	24.5	25.2	25.8	23.4	24.5	25.6	26.5	26.7	24.4	25.2
27.7	30.4	28.6	27.6	29.9	28.8	29.5	30.1	27.4	28.5	30	30.7	32.3	29.5	29.3
27	30.1	28.5	27.3	29.7	28.7	29.4	30.3	27.4	28.3	30	30.7	31.3	29	28.6
27.6	30.6	28.2	27.2	29.8	28.8	29.7	30.7	27.5	28.6	30.4	31.5	32.3	29.2	28.6
26.5	30.2	27.5	27.2	29.7	28.8	29.7	30.8	27.4	28.8	29.4	29.3	31.9	28.6	27.9
27.1	29.2	28.1	25.9	29.9	28.4	28.7	29.2	26.9	27.4	29.6	29.8	30.4	28	29.1
27.1	29.4	27.2	26.7	30.1	28.9	29	30.6	27.3	28.2	30.5	30.5	31.2	28	28.1
29.4	32.4	30.7	28.4	31.6	30.7	31	32	29.3	29.8	31.6	32.1	32.9	29.9	30.4
26	27.1	24.5	25.2	25.6	25.4	21.6	24.2	23.9	24.9	23.8	21.9	27.4	25	22.7
28.8	27.3	29	28.8	27.3	27.3	26.1	25.8	27	28	25	25.4	29.9	29.4	27.5
22.7	20.2	19.9	21.7	20.8	20.7	17.6	18.7	20.5	21.1	16.9	18.6	23.7	22.9	20.9
33.8	30.9	30.5	31	32.7	32.2	30.6	32.1	31.6	31.7	29.9	29.7	34.4	34.3	31.7
27.5	25.1	26.3	28	25.7	25.9	23.5	23.2	25.8	26.9	22.9	22.7	28.1	28.4	26.2
31.1	29.4	29.7	30.2	29.6	27.8	27.5	28.5	29.6	30.2	27.5	27.6	31.3	31.6	29.7
23.1	22.4	23.1	23.4	22.4	23.3	20.5	21.8	22.9	24	19.5	20.2	25.4	25.4	23.5
25.7	24.8	25.1	24.5	24.5	24.7	22.7	23.9	24.7	26	22.4	22.7	27.4	27.4	25.6
27.1	25.5	25.9	26.5	26.6	27	24.5	24.3	25.2	28.7	24.2	24.1	28.9	28.7	26.5
26.7	24.1	25.2	25	24.8	26.1	21.9	21.7	25.1	25.8	22.9	22.7	27.4	27.3	25.4
28.6	27.6	27.8	28.3	29.1	27.7	27.2	27.5	28.1	28	28	29.1	29.2	29.8	25.2
31.6	29.7	27.5	29.8	32.4	29.5	29.1	32.7	31.1	31.2	28.6	28.5	32.8	32.4	29
31	29.2	28.1	29.7	30	28.2	27.2	27.5	28.1	30.1	26.7	26.6	31.4	31	28.5
26.4	24.9	25.6	25.7	25.6	25.8	25	24.6	25.3	25.8	22.5	23.5	28.4	27.8	25.3
29.2	28.5	29.5	28.9	29.4	27.1	27.7	27.5	29	28.5	26.9	27.4	29.3	29.5	25.6
35.2	33	31.7	33.3	33.5	32.6	31.4	33.3	32.2	32.3	31	31.2	34.9	35.1	32
29.3	28.2	28.9	28.6	28.4	27.5	27.2	27.2	27.5	27.6	25.7	26	28	27.9	25.8
32.5	31.1	30.9	31.2	31.7	31.7	29.6	30.7	31.2	28.7	27.5	26.3	33.3	31	28.2
31.6	31.9	30.1	30.8	32.7	30.5	31.1	31.4	31.3	32.1	31	32.6	32.3	33.8	30.1
31.3	29.9	31.1	31	32.3	29.9	29	29.7	30.2	31.1	30.1	31.2	32.2	31.3	26.3
29	30.3	27.5	28.5	30.9	30.2	29.3	32.1	30.5	32.3	30	30.7	32.6	32.1	29.7
29.4	29.1	28.6	28.5	30.7	28.9	29.3	30.3	28.8	29.5	28.7	29.9	29.5	30.9	26.9
33.1	30.8	30.9	31.3	31.8	31.4	30.2	30.2	28.7	31.2	29.4	29.6	32.6	32.5	29.6
31.9	30.5	30	31	32.7	32	30.4	31.8	30.5	31.4	30	29.1	34	32.5	30.2
30.9	30.5	30.9	30.4	32.3	29.6	29.9	30.5	30	30.3	30.6	30.8	30.8	32.4	27.3
31.2	30.3	29.3	29.4	30.2	30	28.7	28.8	29.2	29.2	27.9	28.2	32.1	30.5	28.3
30.6	30.5	30.8	31.3	32.6	31.2	31.2	30.7	31.4	30.9	30.1	30.9	31.1	32.7	28.8
29.4	29	27.8	28.6	28.1	26.9	27.4	25.8	28.1	28.7	26.5	27.8	30.2	30.3	27.1
32.5	32.3	30.5	31.1	33.2	31.6	30.7	33.1	31.5	32.5	30.4	30.9	31.8	31.9	28.8

1991	1992	1993	1994	1995	1996	1997	1998	1999	2000	2001	2002	2003	2004	2005
34.3	32.2	35.5	30.5	33.3	32	33.1	32.3	33.9	34	32.1	32.7	30.6	31.5	32.7
32.7	31.6	34.8	30.2	34.9	32.2	32.1	33.2	32.3	34.6	32.3	32.5	31.1	31.8	32.2
33.3	32.8	34.3	30.5	33.4	32.8	32.6	33.4	33.2	34.9	33.3	32	31.2	32.2	32.1
32.6	33.1	34	31.4	33.8	33	32.4	34.9	32.7	36.5	33	34	32.4	32.9	33.3
32.6	32.3	32.5	30.9	32.5	32.6	32.1	34	31.9	34.7	31.7	32.5	30.9	33	32.4
33.4	32.5	34.6	30.3	34.2	33	32.8	34.1	32.5	36.3	32.7	33.3	31	32.3	32.6
33.8	32.9	35.2	31.2	34.2	33	33.4	34	33.9	36.4	33.5	33.8	31.7	32.6	32.5
34	33.6	35.3	31.7	35.2	33.8	33.6	34.8	33.6	37.2	33.8	33.9	32.5	33.4	33.3
32.3	31.1	34.5	29.7	32.8	31.2	32.4	31.9	31.9	33.2	30.6	32.2	30.4	30.6	31.3
32.4	32.2	33.9	30.8	33.8	33.4	32.4	33.5	32.8	35.8	32.5	33	31.5	32.2	32.1
32.6	32.2	34.7	30.9	34	32.2	32	34.3	32.8	35.2	32.7	32.8	30.9	32	32.3
31.4	31.8	34	29.6	33.2	32.2	31.8	32.7	30.9	35.2	31.4	31.2	29.7	31.2	31.5
30.2	30	32.8	27.8	31.3	29.6	30.8	30.6	31	30.6	29.1	30.5	28.5	28.7	29.3
32.1	31.1	34.6	29.5	32.9	30.6	31.5	31.7	32.3	33.6	31.9	31.3	30.3	30.5	30.8
33.8	33.9	34.4	31.2	34.5	33	32.2	34.1	32.1	35.9	32.3	33.5	31.2	32.4	32.2
34.7	31.2	32.6	31.9	32.1	31.3	33	35.1	33	31.7	34.5	32.7	33.4	29.5	33
33	31.6	34.6	30.7	32.7	32.3	33.2	35.9	34	33.4	34.1	32.7	31.7	31.4	33
32.7	30.3	32.3	30.1	31	30.1	32	33.8	31.5	30.6	32.3	30.8	30.9	28.3	31.9
31.8	30.2	34	30.2	31.7	30.5	32.6	36	32	32.1	33.7	31.2	31.7	29.8	32.4
33.7	32.2	34.9	31.5	32.8	31.6	34	35.5	33.7	33.2	34.7	32.3	33.8	30.5	33.7
33.9	32.2	34.5	31	32.5	31.2	34	35.5	33.4	32.8	33.8	31.5	33.5	30.4	32.8
33.5	31.9	35.2	32.7	33.6	32.5	33.9	37.2	33.9	33.4	33.5	32.5	31.9	30.4	33.1
32.9	32.6	35.9	31.6	33.9	32.3	34.2	35.6	34.1	33.9	33.7	33.6	32.5	32.5	32.9
33.9	31.8	34.7	31.5	32.1	32.5	34.4	32.6	33.8	31.7	34.1	32.2	32.9	30.6	32
33.4	31.8	34.9	29.2	32.1	31.8	33.2	34	32.8	32.8	32.5	32.3	31.8	30.4	31.9
33.2	31.1	31.3	30.5	31.9	30.3	31.7	34.1	31.9	30.4	32.5	30.3	31.7	28.8	32.5
33.9	31.9	34.8	31.5	31.9	31.8	33.5	32.8	34.3	32	33.2	32.8	32.8	30.8	32
32.7	31.1	34.5	30.1	31.7	31.7	33.3	33	32.6	31.8	32.8	32.5	31.4	30.3	32.1
34.2	31.7	34.8	30.6	32.2	32.2	33.7	32.5	33.8	32	33.1	32.1	31.7	30.4	32.2
33.4	32.6	35.2	31.8	32.9	32.6	33.7	34.7	32.9	33	32.6	32.6	31.7	30.9	31.7
31.7	30	32.6	33.6	32.5	33.7	31.6	33.6	29.9	34	32.6	34.5	36	33.3	35.4
31.7	30	32.3	33.1	32.6	33.3	31.4	31.6	29.6	32.9	31.7	32.8	35.9	32.2	35
37	36.4	37	38.4	38.9	38.6	38.7	38	35.3	39.2	37.3	39	40.1	38.7	39.7
32.3	32	31.1	35	34	32.9	34	32.5	30.2	34.1	31.9	33.2	35.7	33.2	36.5
38.3	37.2	37	39.4	40.1	40.4	39.3	40.2	37.1	39.8	38.2	39.9	40.9	39.2	40.5
28.2	26.2	28.3	29.7	28.2	28.5	28.5	28.9	25.5	29	26.9	30.6	31	28.3	30.9
32.5	32.1	32	33.7	33.2	33.1	32.9	33	29.7	33.9	32.5	34.5	35.2	33	34.6
36.2	37	36.6	38.8	38.6	37.7	39.2	36.3	34.7	38.2	36.1	37.2	39	37.3	39.5
26.4	24.9	25.9	27.7	27.3	27.1	27.6	27.3	24.5	27.9	26.8	28.4	27.4	27.5	29.4
38.9	38.5	38.8	39.5	40.7	40.4	40.4	40.3	38.1	41.8	39.6	39.9	41.5	39.7	41.8
37.8	35.9	36.7	39	37.5	39.3	37.1	38.2	36.3	39.1	37.4	40.2	41.1	38.1	39.9
31.6	30.8	30.7	32.5	32.6	32.2	32.8	32.1	29.4	32.5	31.8	33.1	34.4	32	33.8
27.9	26.5	27.8	28.9	28.8	28	28.3	28.3	25.1	29.3	27.2	29.5	30.6	28.4	29.8
39.4	40.5	39.8	41.8	41.6	41.9	41.3	42	40.3	41.3	41.2	41.8	42.3	41.4	43.2
41.2	40.7	39.8	43	42.5	42.5	41.2	43	40.3	42.7	41.3	43.3	43.5	42.4	43.7
32.5	32.1	32.8	33.8	34.4	33.8	34.7	33.1	30.2	33.7	32	33.5	36.3	33.7	35.7

39.3	39.7	39.7	41.2	41.5	40.6	40.4	41.4	39	41.4	39.6	40.8	41.8	40.4	42.2
34.3	31.6	33.2	35	34.2	34.7	33.1	34.2	32	35.1	33.3	35.6	37.3	33.9	36.1
40.6	40.6	40.1	42	42.4	42.7	42	42.9	39.4	42.4	40.7	42.5	43.2	41.2	43.2
37.9	37	38.2	39.2	40.1	39.4	39.6	38.3	36.8	40.8	39.3	40.4	41.8	38.8	40.7
35.2	34.1	33.9	36	36	36.1	35.1	36	33	36	34.8	36.6	37	35.4	37.1
36.5	35.4	36.2	37.9	37.6	37.6	37	37.3	34.4	37.9	36	37.1	39.7	36.9	39.5
40	38.6	40.4	41.4	41.5	41.1	40.7	40.9	38.7	41.2	39.8	41.5	41.8	40.2	42
32.9	31.3	32.1	34.1	33.4	33.5	32.7	32.6	29.7	33.3	32.3	32.8	35.3	32.3	34.6
37.2	36.3	36.5	38.5	38.7	38.1	37.9	37.5	34.2	38.2	36.9	37.7	39.5	37.1	39.5
34.4	32.3	30.4	34.3	31.3	35.6	33	33.8	32.8	31.4	34	35.6	35.7	35	34.9
34.5	32.7	29.8	36.4	30.1	35.6	32.5	34.4	33	32.2	33.7	36.1	36.2	34.8	35.6
30.1	27.6	27.2	31.8	29.1	32.3	30.4	29.3	28.7	30.1	31.5	32.9	34.4	32.4	34
32.5	29.4	27	34	29.6	32.9	30.2	32.5	31.1	31.3	31.9	34.4	34.8	32.7	34.2



30.6	30.2	32.1	31.6	31.5	28.8	30.7	30.5	32.4	27.9	28.8	32	29.9	30	31.4
29.8	30	32	30.5	30.9	28.5	30.1	30.2	31.3	27.9	28.1	31.6	29.4	29.4	30.6
32	30.7	32.5	32.4	32.4	33.2	33.1	33.4	32.4	32.6	31.7	32.8	32.1	32.7	33.4
33.4	34.1	34.9	31.8	34.6	33.7	33.4	34.9	33.6	35.7	33.2	33.7	31.8	33.6	33.6
31.8	32.3	32.3	31.2	31.2	31.7	31.9	32.9	31.8	32.2	30.8	30.9	31.6	32.2	32.2
33.8	33.9	33.6	32.9	33.2	33.8	34	34.1	33.1	33.4	33	33.1	33.8	34.2	34.2
32.4	31.6	32.4	31.5	31.9	31.9	31.9	32.4	31.5	32.3	31.3	31.8	31.9	31.9	31.9
31	32.8	32.1	31.3	32.2	32.1	32	32.7	32.3	31.5	31	31.1	31.8	31.6	32.7
31.7	33.1	33.3	30.4	32.6	31.3	31.2	32.8	32.1	32.6	31	31.3	30.1	31.7	32
32.4	32.4	32.5	31.9	32.4	32.6	32.6	33.2	32.2	33.1	32	32.1	32.1	33.1	32.9
31.6	33	32.4	32.1	31.8	31.8	32.4	33.3	32.2	32.3	31.1	31.3	31.8	32.3	32.4
33.8	33	34.7	31.8	34.3	34.8	33.5	34.7	33	36	32.4	33.8	31.9	33.6	33.5
32.8	34	32.7	32.6	32.6	33.4	33.5	34.2	33.2	33	32.8	32.7	33.5	33.1	33.7
31.8	33.1	33.7	31.7	33.1	32	32.5	33.3	32.6	33.1	32	31.9	31	32.4	32.7
31.8	33.1	33.7	31.7	32.6	32.4	33	32.8	32.8	32.4	32.6	32.4	32	33.2	33.1
31.4	32.5	32.8	31.4	31.7	32.3	32.4	32.9	31.9	32.4	31.4	31.3	32.1	32.2	32.2
31.6	32.7	33.6	31.6	32.1	31.7	32	35.5	32	32	30.9	31.2	31.6	31.6	32.1
32.3	33.9	33.5	32.5	33	33.3	33.4	33.5	33.2	33.1	32.2	32.9	33.1	32.6	33.3
33	33.5	34.5	31.9	33.4	32.8	32.5	33.7	33.5	33.8	33.5	32.3	32	33	33.7
32.5	33.4	34.5	31.8	34	32.6	33.5	34.3	33.6	34.2	32.8	33	31.2	33	33.4
32.2	32.2	33.1	30.5	32.8	32.3	31.8	33.5	31.9	34.6	31.9	32.6	30.5	32.1	32
32.6	33.6	33.2	33.2	32.4	33	33.7	33.2	33	32.6	31.8	32.2	33.3	33.1	33.5
32.3	35	33.8	32.8	33.6	33.3	33.7	34	33.8	32.8	32.6	32.5	32.9	32.6	33
32.4	33.8	34.3	32.6	33.6	33.5	33.1	33.7	33.7	32.9	32.4	32	32.4	32.8	33.6
30.2	30	34.4	28.1	32	30.2	30.7	31.7	30.9	30.9	29.5	31.8	29.2	29.8	29.9
33	32.8	38	30.7	34.5	33.2	34.3	35.1	33	34.7	32.3	33.9	31	33.4	32.7
33.2	34.2	36.4	32.5	33.8	34.1	32.9	34	33.4	34.3	32.2	33.5	32.7	33	33.8
31.4	31.4	36.2	29.9	33.5	31.2	31.8	32.9	32.4	33.6	31.7	32.9	30.6	31.6	31.4
32.4	33.3	34	30.8	33.4	33.3	32.1	33.6	32.1	33.7	32	33.8	31.2	33	33
32.3	32.9	36	31.5	34.7	34	33.7	35.7	34.2	34.8	32.5	35.1	31.8	34.4	32.1
32.6	33.9	34.9	32.2	33.4	33.2	33.2	34	33.6	33.2	32.5	33.9	32.8	33	32.6
32.3	32.3	35.4	31.1	33.4	33	32.4	35.5	32.6	33.7	31	34.1	31.2	32.9	32.8
34.1	34.6	35.7	32.9	34.6	33.8	34.6	35	34.2	34.6	32.8	34.2	33	35.2	34.2
33.8	34.2	36.6	32.7	34.5	34.6	34.2	36	34	35.7	33.2	34.9	32.7	34.5	34.5
33	33.1	34.6	30.8	34.6	33.4	33.3	34.7	33.1	34.8	33.3	33.3	30.7	32.7	33
32.2	33	33.6	30.3	33.5	33.2	32.9	34.7	32.7	34	32.4	34	32.1	32.4	33.1
31.2	31.5	35.4	29.6	32.8	31.5	32.3	32.9	32.1	32.3	29.9	32.9	29.8	30.8	30.9
31	31.1	35.9	29.4	31.3	32.3	32	32.5	31.5	32.2	30.5	32.4	30.2	31.5	31.2
32.5	32.1	35.4	30.5	33.8	33.4	33	34.2	32.9	33.9	32	33.7	30.4	33.8	32.1
32.7	33.8	34.7	31.2	34.4	34.2	33.8	34.9	33	34.9	33.5	34.5	31.9	33.4	33.2
29.8	29.9	32.9	28	32.7	31.4	30	31.9	30.9	32.7	30.5	31.2	29.6	30.9	30.5
31.8	32.8	36.1	30.4	35.3	32.1	32.5	34.3	32.9	33.8	31.1	33.6	30.6	32.4	31.9
30.3	30	33.9	28.4	32.4	29.8	30.4	32	30.7	31	29.8	31.6	29.1	30.1	30.1
32.9	31.9	34.4	30	34.9	33.2	32.3	33.1	32.1	34.8	33.4	32.7	31	32.1	32
31.3	32	32.5	29.4	32.2	32.1	31.8	32.8	31.8	33.5	31.7	32.2	30.2	31.4	31.4
31.1	31.9	34.8	29.1	33.6	31.8	31.7	33.5	32.1	33.9	31.4	32.1	30.1	31.2	31.1
31.2	31.7	36.2	30.2	32.9	31.9	32.3	33.7	32.1	33.2	30.5	32.9	29.6	31.3	31.5

27.6	23.8	25.2	25.7	28.5	25.5	27.4	28	28.1	26.5	28.1	30.2	27.7	26.2	28.7
27.7	22.7	25.7	25.7	28.1	25.8	27.7	27.4	29.4	26.7	28.9	30.5	27.7	26.1	28.6
29.8	25.6	26.8	27.9	29.2	28	29.4	28.8	32	27.6	30.5	30.8	28.7	27.5	31.2
30	25.5	27.5	27.2	29.4	27.5	28.3	28.8	30.9	27.3	29.6	31	28.4	26.6	29.6
28.4	23.7	25.5	25.1	28.1	26	26.4	27.5	29.7	26.6	28.5	29.2	27.1	25.3	28
29.8	26.9	27.8	28.8	29.8	28	30	29.5	31.1	28	29.8	31	29.2	27.1	30.5
29.4	26.5	27.7	28.1	29	27.4	29.8	28.5	29.7	27.7	29.1	30.5	29.9	27.8	29.8
28.1	23.6	26.3	26.1	29.1	26.4	28.4	28.1	30.4	27.2	28.8	30.3	27.9	26.2	28.6
26.9	23.2	25.2	24.9	27.4	25.5	25.9	27	29	26.2	27.8	29	26.8	25.5	28
30.1	24.9	27.5	26.7	28.5	27	28.2	28.4	30.5	26.9	29.2	30	28	26.4	29.1
31.4	26.9	28.1	29	30.7	29.6	29.6	29.7	31.7	28	30.4	31.2	29	27.8	31.1
28.8	24	26.5	26.7	29.5	27.7	29.4	28.9	30.6	29.1	30.1	32.7	30.4	28.5	29.9
27.5	23.7	25.7	25.5	28.1	25.9	28.8	28.3	30.1	26.6	28.7	30	27.9	26.4	29
29.8	25.5	26.7	27.4	28.8	27.8	28.4	28.1	31	26.2	30.1	30.6	28.5	25.9	30.1
28.5	25.3	27.1	27.2	30.4	27.7	29.1	28.8	30.2	29.4	30.2	31.2	29.7	27.7	29.8
28	23.8	25.4	25.4	27.7	26.1	26.3	27.7	29.3	27	29.6	29.5	27.4	25.5	28.2
30	25.9	27.7	28	30.3	27.5	29	28.8	31	27.3	29.6	30.5	28.6	26.9	30.2
28.9	24.8	26.4	26.9	29.8	27.4	28.9	28.6	30.2	27.5	29	30.4	28.1	26.4	29.2
30.2	25.4	26.8	27.1	29.1	27.8	28.9	29.3	30.8	28.1	29.6	30.4	28.5	27	29.5
29.8	26.4	27.4	28.6	29.7	28.1	29.1	29.2	31.7	27.9	29.8	30.6	29.6	26.8	30.2
29.7	26.5	27.5	27.9	30.8	28.1	30	30.3	31.1	29	30.7	32.3	30.5	27.5	31.2
29.3	24.9	26.1	26.9	29	27.2	28.9	28.5	30.5	27.4	28.8	29.7	27.6	26.2	29.3
30.4	28.1	28.5	29	31.4	28.3	30	29.5	31.2	28.6	30.7	32.3	31.5	27.6	30.5
31.7	28.8	24.1	33.3	29.3	32.8	29.5	32.6	32.5	33.1	32.2	35	35	32.9	33.4
32.3	28.6	25.4	33.9	30.2	33.7	29.2	33.7	30	32.6	31	33.7	35.7	33.7	32.8
31	28.8	24.3	31.8	29.1	31.1	27.9	32	30.6	32.1	31.6	33.5	34.9	31.2	32.6
27.5	24.4	21.2	28	25.4	28.6	25.2	29.3	28.2	29.9	29.2	30.3	31.8	27.6	28
27.6	25	20.7	29	27.4	28.3	27.2	30	26.2	27.1	26.6	28.7	30.7	28.8	27.5
29	27.6	23.1	31.1	28.4	29.6	27.8	31.3	27	30.1	29.1	30	33.1	29.9	30.2
28.8	26.3	23.4	29.6	26	28.4	25.6	29.3	27.8	30.2	29.6	31.1	33	28.9	30.4
27.6	25.4	20.4	30.2	26.8	28.5	25.9	29.7	26.4	27.2	27.5	28.9	31.1	28.8	28.3
28.4	24.6	20.7	28.5	25.4	27.5	25.3	29.1	26.3	27.8	28	29.4	30.4	27.9	28.7
26.6	24.2	21.4	27.7	24.9	27.3	23.9	27.1	26.7	28.4	28.7	29.9	30.8	26.9	28.7
29	26.3	21.4	31	27.1	29.4	25.9	30.2	27.6	28.3	27.6	30.8	32.4	29.8	29.9
26.4	24.3	20.1	29.1	25.8	27.9	25.3	29.4	25	26.3	26.3	27.6	30.4	27.9	27.5
29.9	27.7	23.5	30.6	28.4	30.5	27.2	31.7	29.6	31.6	30.2	31.8	33.2	30.3	31.5
33.5	30.1	25.8	34.3	31.1	33.7	31.5	33.8	31.6	32.8	31.3	34.5	35.7	33.3	34.7
33	29.9	25.3	32.3	29.7	32	29	32.9	30.8	32.9	31.1	33.7	34.8	31	33.6
31.4	29.7	24.9	33.3	30.5	32.9	29.4	34.2	29.6	30.4	30.3	34.6	34.2	32.8	32
34	31	26.8	34.5	32.1	34.6	31.4	35.2	32.5	33.4	32.9	35.1	36.6	34.2	34.8
37	33.9	29.2	37.3	33.8	36.6	32.5	37.3	36.1	37.1	34.9	37.7	39.4	36.7	37.2
32.6	28.9	24.5	32.6	30	32.8	28.5	33.3	30.9	32.5	31.1	33.4	35	30.7	32.9
33.8	29.6	26.6	33.8	31.7	34.9	30	35.7	31.6	34	31.8	34.6	35.7	34	35.2
29.7	28.1	23.5	30.7	28	29.4	27.8	30.7	29.7	30.2	28.5	31.9	32.8	30.3	31.4
26.7	24.2	19.9	27.5	24.1	27.7	24.3	28.2	25.5	26	24.7	28.7	29.3	26	27.5
27.7	27	22.3	31.1	27.2	29.6	25.8	30.3	27.1	27.7	27	29.3	31.6	29.7	29.8
31	26.6	23.5	30.5	28.5	31.5	27.5	31.2	30.3	31.9	29.1	32	33.1	29.7	31.5

31.4	26.2	22.8	29.4	27.9	31.9	28.9	31.1	30.4	31.9	28.9	32.1	33.7	29.3	31.7
30.7	28.2	23.4	31	28.2	30.2	27.4	31.5	29.7	30.5	29	31.9	33.4	30	31
32	30.3	25.4	34.3	31.4	33.4	30.2	34.6	31	31.9	31.1	34	35.7	33.8	33.5
29.9	26.9	23	29.5	26.3	29.4	26.4	29.9	28.6	29.7	28	31	32	27.5	30.5
30.1	27.9	22.2	32.8	28.8	31.5	27.7	32.5	27.4	28.3	28	31	32.7	30.4	29
30.6	28.8	30.4	29.5	31.1	29.7	30.5	29.3	32.3	28.2	30.3	32.3	30.2	27.8	31
31.4	29.7	31.4	30.8	32	30.2	30.8	29.6	32.6	28.9	29	32.8	30.8	29	31.2
29.3	24.7	27.1	26.8	29.1	26.9	28.4	28.8	30.5	27.2	29.4	30	28	26.9	30
29.7	27.4	28.6	28.5	29.2	28.4	30.1	29.3	31.5	28	30	30.9	29	27.6	31
29.3	24.9	27	26.9	28.7	26.7	28.2	27.9	29.8	26.6	28.9	30.1	28.3	26.6	30.1
30.2	25.1	27.3	27.2	29.4	27.4	28.9	29.1	30.8	27.5	29	30.3	28.3	27	29.8
31	30.2	32.9	30.2	31.6	29.9	31.9	30.2	32.4	28.7	30.8	32.2	31	29.5	31.3
30.5	29.1	29.7	29.4	30.6	29.1	30.9	29.9	32.2	28.9	30.8	31.6	30	28.5	32.2
31.1	27.5	29.3	29.8	31.2	28.6	29.6	29	31.4	28.3	30.3	32.2	29.6	27.9	30.5
32.4	31.4	33	31.3	32	30.2	31.8	30.6	32.9	30	31.4	33.5	31.4	29.6	31.9
29.3	25.6	26.8	27.2	29.1	27	28.3	28.1	29.9	26.2	28.4	29.7	28	26.5	30.5
30.9	26.1	28.1	27.9	30.2	26.6	28.9	28.8	31	27	30.3	31	28.4	27.4	30.1
30.6	28.2	30.2	29.9	31	28.9	30.4	28.9	31.3	27.7	29.9	31.6	29.2	27.5	30
30.2	28.8	29.8	29	30.4	28.4	31.6	29.8	32.3	28.3	30.4	31.6	30	28.3	30.7
30.6	28	30.1	29	30.1	29	30.5	28.7	32.2	28.2	30	32.6	29.8	28.4	31.3
30.7	25.5	28.3	28	30.7	27.5	29.5	29	31.3	27.8	30.7	31.3	28.1	27.1	29.9
31.1	31	31.4	30.4	30.9	29.5	31.5	30.1	32	29	30.6	33	30.5	28.3	31.6
29.4	25.4	27.1	26.9	29.4	27.1	28.3	28.6	31	27.1	29.5	30.3	28.4	26.8	30
31.9	30.1	32.1	29.4	31.1	29.5	30.9	30.5	32.2	29.2	30.8	33	30.8	29	31.4
29.7	27.7	29.7	28.5	29.5	28.3	29.6	28.1	31	27.4	30	30.6	29.2	27.3	29.7
29.9	28.6	30.5	28.1	28.4	27.9	29.9	28.8	30.9	28.1	29.3	30.6	29.3	27.6	29
31.9	31.4	32.9	31	32	30.2	31.7	30.6	33.4	30	31.2	33.5	31.9	30.2	31.5
29.9	29	29.7	29.3	30.4	29.5	30.8	29.1	31.6	28	29.7	31.2	29.5	27.7	31.1
30.2	26.2	28	29	30	28	29.1	28.3	30.5	27.2	29.4	30.3	28.4	27.3	30
33	29	33	30.7	31.5	29.9	31.5	30.3	32.6	29.9	30.5	32.9	31	30.1	31.1
31	26.6	28.5	28.5	30.6	28.6	28.5	28.6	30.6	27.4	29.2	30.4	28.1	27.1	29.5
32.1	27.7	29.2	29.2	31.5	29.5	30.9	30.4	32.7	28.5	29.9	31.3	29.2	28.3	31.9
29.6	25.8	26.9	27.5	29.2	27.8	28.9	29.5	30.7	27.4	29.5	30	28.7	27.1	30.5
30.6	27.3	28.8	28.5	29.8	27.8	29.3	28.1	30.7	28	29.3	31	28.4	27.1	30
29.9	27.4	28.1	28.4	29.7	27.8	29.3	29.7	31.2	28.3	30.2	30.7	28.8	27.1	30.6
30.6	28.2	29.4	29.6	30.6	28.8	29.3	29.1	31.2	27.8	30	31.3	29.1	27.8	30.4
30.8	27	29.1	28.7	30.1	28.1	29	28.3	30.8	27.8	29.7	31.4	29.1	27.2	29.5
29.3	24.8	26.6	26.7	29.2	26.7	28.6	27.4	32	27.8	30.3	30.5	28.2	26.8	29.9
29.3	25.1	27.7	27.4	29.7	27.4	29.6	28.8	31.2	27.4	29.3	30.9	28.7	28	30.4
29.3	28.1	29.2	28.4	29.4	28.4	29.5	29.2	31.2	27.6	29.8	31	28.8	27.5	30.8
32.3	30.6	32.1	30.2	31.3	30.2	31.5	30.6	32.1	29.1	30.9	32.4	30.9	28.7	31
29.2	25.6	27.4	27	28.7	26.7	27.3	27.7	30.6	26.4	28.2	30.4	27	26.4	29.6
31.1	27.9	30.7	29.3	29.8	28.5	30.1	28.7	32	27.8	29.2	31.1	29.2	28.4	29.6
31.9	30	32.2	30.1	30.7	28.2	31.2	29.2	32.8	28.5	30.4	32	29.7	29.1	30.5
32	29.5	32.1	30.1	31.3	29.2	31.6	29.8	34.2	30	31	32.3	30.2	29.7	31.8
32.6	29.9	32	30.4	31.7	29.5	32.3	30.3	33.9	29.9	30.8	32.5	30.5	29.6	32.1
30.6	27.1	29.3	28.6	29.5	27.6	28.3	28	31.3	27.3	28.8	30.9	28.1	27	28.8

28.4	25.4	27.4	27.5	28	26.5	27	28.1	29.9	25.7	27.6	29.9	26.6	25.8	28.7
29.3	27.5	29.4	28.4	29.9	27.8	29.2	28.1	30.6	27.5	29.3	30.6	27.8	27.1	28.9
29.8	26.2	27.9	27.5	28.9	27	27.8	27.9	30.7	26.6	28.3	30.6	27.9	27.1	29.5
32.8	29.8	32	30.6	30.9	29.9	31.6	30.3	32.7	29.6	30.6	33.3	30.1	29.7	31.6
31.7	29.6	31.4	29.9	31.1	29	31	29.3	32.5	28.8	30.2	31.5	29.9	28.4	30.1
30.6	28	29.8	28.9	30.1	28.2	29.9	28.7	31.4	27.7	29.2	30.8	28.7	27.4	29.6
29.5	25.5	27.4	27.5	29.4	27.1	28	28.1	30.6	26.9	28.6	30.1	27.6	26.6	29.4
30.7	27.3	28	28	30.3	27.8	28.9	28.8	30.8	27.9	29.4	31	28.3	27.5	30
33.8	30.9	32.9	31.4	31.7	30.4	32.2	30.4	32.9	30.2	31.3	33.6	31.1	29.4	31.5
29.7	26.4	28.5	27.4	29.1	26.8	28	27.3	30	26.6	27.9	29.4	27.2	26.3	28.9
29.4	26.5	28.5	27.7	28.8	27.1	28.1	27.6	30	26.6	27.6	30.1	26.8	26.3	28.5
33.3	30.5	32.4	30.9	31.9	30	31.5	30	33.6	29.5	31.5	34	31.5	29.2	31
32.2	30.7	32.9	31.3	31.6	29.9	31.3	29.8	32.9	28.9	31.1	33.4	31.2	29.5	31.1
30.7	28	30.6	29.4	30	28.1	29.9	28.5	31.7	27.9	29.2	31.1	29	28	29.9
30.2	27.1	29.1	28.9	29.8	28.3	29	28.4	31.8	27.6	29.1	31.4	27.9	28	29.3
31.2	29.8	32	31.2	31.5	29.9	31.7	30.1	33	29.3	31	32	29.8	29.5	30.6
29.4	26.5	29	28.4	29	27.1	27.9	28	32.2	27.2	28.7	31.4	27.7	27.2	29.1
31.3	26.8	28.7	27.5	31	27.6	29.2	29	31.8	28	29.7	31.3	28.8	27.1	29.1
31.3	28.8	31.1	29.5	30	28.4	30.3	28.8	32	28.4	29.7	31.6	29	29.3	29.9
30.4	27.8	29.8	28.9	30.1	28.4	29.6	28.9	31.8	28	30.3	30.7	28.5	27.5	29.3
30.6	27.3	29.3	28.6	29.8	28.2	28.8	29.4	32.7	28	28.9	32.1	28.1	27.6	30.2
30.6	26.4	28.7	28.1	30.5	27.6	28.6	28.6	30.9	27.5	29	30.6	27.9	27	29.9
29.4	25.5	28.1	26.8	29.5	26.6	27.6	28.1	30.9	26.3	28.3	30	27.2	26.3	29
32.2	30.6	33.1	31.4	31.6	30.5	31.6	30.4	32.4	29.5	31.3	33.4	31.7	29.4	31.1
31.8	29.4	32	30	30.9	29.4	31.6	29.5	32.3	29.1	30.4	32.3	30	28.9	30.5
29.4	25.1	27.6	26.9	28.9	26	27.6	27.6	30.3	25.8	27.9	29.9	27.2	26.4	29.3
31.5	28.8	30.7	29.1	29.8	28.4	30.1	29	31.7	28.4	29.7	31.6	29.1	27.8	29.5
31.6	29.2	31.5	30.3	31	29.3	30.7	29.1	32.4	29.1	30.2	32.3	29.8	28.7	30.4
30.4	26.9	29.3	28.2	29.4	27.3	29.1	28.3	30.4	26.9	28.9	30.9	28	26.8	28.9
30.1	27.5	29.7	28.2	29.2	27.9	29.5	27.6	30.5	26.7	28	29.8	27.7	26.3	28.9
34.7	31.7	33.6	33.1	33.8	32.5	33.5	35.2	34.4	34.4	37.2	36.6	38.3	32.1	35.1
33.8	31.6	33.2	33.6	33.8	32.3	34.3	34.3	35.5	35.7	37.6	35.4	37	32	34.6
35.6	30.5	31.5	32.5	33	32	33.2	33.2	34.3	34.1	36.4	35.5	36.9	31.4	35
33.1	29.6	29.5	31	32.8	30.5	34.3	33.1	34	34.7	35.5	35.6	35	30.1	33.2
33.8	30.3	31.8	30.7	33	31.4	32	31.8	32.7	32.3	36	34.2	36	30.3	32.6
31.3	30.2	30.7	33	33.3	31.7	33.3	32.7	34.7	34.3	35.8	35.4	36.9	31	34.8
36.1	33.5	35	34.5	34.3	33.5	34.7	35.8	35.1	34.7	38	35	37.8	32.6	34.5
33.4	30.7	31.9	32.9	32.8	32.4	33.1	33.7	34.4	34	36.2	34.8	36.4	31.1	34.3
33.4	32.8	33.2	33.9	33.5	32.3	34.7	34.2	34.5	34.5	37.1	33.9	36.2	31.4	33.4
36.3	30.2	30.7	31.2	33.3	31.4	32.9	32.5	34	32.8	35.9	35.4	36.6	30.3	34.5
32.5	29.5	29.7	30.2	31.8	29.5	31.9	30.4	32.4	30.8	32.4	33.6	33.9	29	32.6
36.5	32	31.4	31.5	33.8	32.8	33.6	32.4	33.9	33.8	37.4	37.3	37.9	31.2	34.2
35.1	29.6	29.8	31.1	34.2	31.2	32.9	32.5	34.7	33.7	35.5	36.3	35.9	31.1	34.2
33.8	29.7	29.8	29.8	31.6	28.9	30.7	30.8	32.4	30.4	32	32.9	33.7	29	31.7
34.5	29.6	29.6	30.5	33	30.8	33.2	33.3	34	32.6	35.4	36.4	36.1	30.3	35.3
35.1	32	32.1	31.2	34.6	32.4	33	32.3	33.9	33.3	36.2	34.7	36.4	31.3	34.2
33.5	30.2	31.6	31.5	32.4	30.8	31.4	32	32.5	31.2	33.4	32.3	32.8	29.2	32.1

33.7	29.4	29.7	30.3	31.9	30.2	32.5	30.6	32.9	30.8	33	34.4	34.3	29.7	32.4
34	29.6	30.5	29.9	32.1	30.4	32.4	30.8	33.2	32.1	34.5	34.5	33.5	29.3	32.3
34.3	31.9	33.2	32.2	33.7	34.7	32.4	34.7	33.7	32.8	36.7	33.4	34.8	30.9	33.7
34.2	31	32.2	31.1	32.5	32.6	32	33.5	33.1	31.9	34.4	32.2	33.3	29.9	32.5
33.8	31.9	32.7	32.8	32.1	31.5	31.9	33.2	32.4	31.3	35.8	32.9	35.6	30.5	32.7
35.4	32.5	31.9	32	34.3	32	31.9	32	33.6	33	35.5	34	35.6	31.9	34.1
34.4	29.5	30.1	30.6	31.6	29.6	31	30.7	33.1	31.4	32.5	33	33.5	29.4	31.5
35.5	33.2	33.9	33	34.3	33.7	33.2	35.5	34.6	34	38.1	34	36.6	31.8	33.9
34.2	31.6	32.5	31.6	32.8	32.4	31.3	32.4	33.3	32.3	35.3	32.5	34.5	30	32.9
34.1	30.4	31.6	31.3	32.3	30.7	31.8	32.2	33.7	31.9	33.5	33.1	34.2	29.1	32.3
33.3	29.2	29.2	29.5	31.7	29.8	30.2	30.6	31.6	31.7	33.6	33	33.3	28.6	30.3
35.2	33.1	33.6	32.7	34.1	32.7	33.3	33.7	33.9	33.1	37.1	33	36.5	31.7	33.3
33.4	30.5	30.3	30.9	32.5	29.1	32.2	30.5	33.3	30.9	32.4	32.8	33.7	28.9	31.7
34.8	29.8	30.6	31	32.7	30.2	31.2	31.5	33.3	32	33.8	33	34.1	29.5	31.5
31.3	30.1	32.3	29.9	31.5	28.3	30.4	30.5	33.5	28.9	30.6	31.5	29.6	29.7	31.4
30.9	28.5	30.6	29	31	28.7	31.3	29.2	33.8	29	30.1	31.7	29.5	28.1	30.6
31	30.1	32.2	29.8	30.9	28.5	31.5	29.5	32.7	28.8	30.2	31.9	29.5	28.9	30.8
31.8	30.3	32.4	31.4	31.8	29.4	31.9	29.5	32.6	29.5	30.5	32.5	30.1	29.2	31.2
29.3	28.8	31.6	28.1	30.6	27.2	29.2	28.2	30	28.4	28.5	30.4	27.4	27.6	28.9
31.4	29.6	31.9	29.9	30.4	28.9	30.6	28.8	31.1	28.8	30.8	31.3	29.6	28	30.5
31.5	30.7	34.2	30.3	32.6	30.3	31.9	30.6	32.3	31.2	29.8	32.2	30	30	31.1
30.9	30	32.8	30.2	31.5	28.9	30.6	29.5	33.9	28.6	29.2	31.1	28.7	28.7	30.9
32	30.6	33.7	30.5	32	30	32.3	30.9	33.1	30.4	31	32.1	30.2	30.3	32
30.6	29.1	31.9	29.1	30.7	28.5	30.4	29.5	32	29.1	29.3	30.4	28.8	28.6	30.4
32.6	32.2	34.4	31.2	33.4	30.4	32.6	31	34.5	32.2	32.2	32.7	30.8	30.4	31.8
31.3	30.5	33.2	29.9	31	29.4	31.1	29.9	32.7	31.2	31.4	32.2	29.9	29.7	30.6
30.8	28.9	31.7	29	30.2	28.2	30.4	29	31.7	29	30.3	30.6	29.3	28.7	30.1
32.6	32.1	32.6	31.5	32.6	32.7	32.3	34	32.2	33.8	32.7	33.4	32	33.2	33.3
33.1	33.5	32.8	31.5	33.5	32.8	33.2	34.3	32.4	33.9	32.7	32.7	32.2	33.2	34
32.4	31.8	31.9	30.6	32.4	31.8	32.6	33	31.8	33	32	32	31.1	31.8	32.2
33.2	32.7	33.2	31.5	33.3	33	33.4	34.1	33	34.1	32.7	33.5	31.7	33.1	33.3
33.1	32.9	33.2	31.7	33.6	33.4	34.2	33.4	32.7	33.3	32.2	32.6	32.1	32.9	33.9
32.6	32.1	32.1	30.5	33.2	33.1	32.8	33.9	32	35.4	32.5	32.6	31.6	32.2	32.5
32.5	32.1	32.2	31.1	32.4	32.2	33	34.3	32.3	33.6	32.1	32.2	31.1	32.4	32.6
32.7	32.2	33.1	31.4	33.3	32.7	33.9	35.1	32.4	34	33.3	33.3	31.8	32.3	33.4
33.6	33	34.5	32.3	33.5	32.6	34.6	37.2	34	35.3	33.9	34	33.2	32.7	33.7
33.8	33.6	34.3	32.3	33.3	33.3	34.6	36.5	34	36	34	34.3	33.4	32.7	34
33.9	32.7	35	33.3	33.9	32.7	35.4	38	34.1	34.8	34.7	34.2	33.7	32.5	34.6
31.6	31.6	31.7	31	32.3	32.5	32.7	33.5	31.7	32.4	32.1	31.8	31	31.7	32.3
31.9	31.6	31.9	31	32.2	31.5	32.6	33.1	31.4	33.3	32	32	31.4	32.3	32.2
33.2	33.1	33.6	32	33.5	33.5	34.2	35.9	33	34.3	33.2	33.2	32	32.5	33.3
32.8	32.2	33.4	31.4	32.6	32.2	34	35.1	33	34.1	33.1	32.8	32.2	31.2	32.6
34.2	33.6	35	33.5	34.8	33.8	35.1	35.6	34.8	36	33.7	34.5	33.6	33.6	35
33.2	32.6	33	31.4	34.1	33.7	33.6	34.5	33.5	35	32.7	33	32.1	32.6	34.1
32.8	32.7	33.8	32	33.5	32.7	33.8	35.1	32.8	34.3	33.2	33.1	32.2	32.7	33
29.3	26.7	29.1	29.5	28.9	26.5	29	28.5	30.8	26.4	27.3	29.5	28.6	26.9	28.6
27.7	25.6	28.6	29.9	29.5	27	28.2	28.5	30.1	25.7	26.7	28.6	28.2	26.1	28.3

27.6	25.9	29.3	29.4	29.1	26.2	27.5	27.6	29.5	25.4	26.5	28.8	27.8	26.5	28.5
28.5	25.8	29.3	29.5	29.5	25.8	27.8	27.8	29.5	25.6	26.9	29.5	28.1	26.7	28.6
27.9	26.9	29.6	29.8	29.9	27.1	28.5	28.8	30	26.3	27	28.8	29.1	26.8	28.8
26.3	23.9	26.7	27.7	27.1	24.6	26.9	25.9	27.6	23.7	24.3	26.5	26.1	24.1	25.8
28.3	26.3	28.8	29.8	29.1	26.8	28.5	28.1	29.8	25.9	27	29.2	28.3	26.2	28.1
28.3	26.6	28.8	29.9	28.8	27.4	28.6	28.4	30.9	26.3	27.1	28.8	28.3	26.7	28.2
27.9	25.3	28.4	29.7	29.2	26	28.6	27.7	29.2	25.2	25.9	28	27.9	25.4	27.5
28.5	25.6	29	30.3	29.6	27	28.7	28.6	30.7	26.4	27.4	29.5	28.9	26.9	28.5
28.8	26	29.1	29.7	29.6	27.2	29.2	28.8	30	25.8	26.5	29.2	28.6	26.5	29.2
29.7	27.1	29.9	31	30	27.3	29.7	29.1	31.1	26.8	27.5	29.4	29.2	26.8	29.1
29.7	29	31	30.8	30.7	27.8	30.2	29.8	31.5	26.9	27.6	32	28.9	29.6	30.3
29.8	29.6	31.1	30.9	30.8	28.9	30.5	30.2	32.4	27.5	28.3	31.6	29.3	29.1	30.4
32.1	29.8	32.1	31.1	31.3	28.7	31.5	30.5	33.9	28.1	29.3	32.2	29.9	29.8	31.2
31.4	29.6	31.5	31.4	30.5	28.3	30.5	29.8	33	27.4	29.2	31.6	29.6	28.9	30.4
31.3	30.1	32.3	30.2	31.9	28.8	30.8	30.9	32.5	28.3	28.7	32.1	29.8	30.1	31.9
30.8	29.4	31.8	30.8	31.4	28.5	30.3	30.2	31.5	27.6	27.9	32.3	30	29.9	30.9
31.8	29.9	32.4	32	32.2	28.5	31.4	31.3	33.3	27.9	28.4	31.7	29.4	29.5	31
32	30.6	32.4	31.3	31.8	28.9	31	32.1	33.7	28.4	29.1	32.5	30.3	30	30.9
32.6	29.1	32.4	30.8	30.9	29	30.5	30.2	34.1	28	28.4	30.9	29	28.5	31.2
32.7	32.6	33.7	33	33.6	30.2	33	32.6	34.3	29.8	30.5	33.7	31.5	32	33.1
31.2	30.7	33	31.1	31.8	28.5	30.8	30.8	32.5	27.5	28.1	31.1	29.5	29.2	30.6
31.7	30.4	32.4	31.5	32.1	28.8	30.7	30.8	32.5	27.9	28.7	32.2	30.2	30	31.1
31.5	30.5	32.5	31.9	32.3	29	31.4	31.2	33	27.9	28.9	32.3	29.9	30	31.3
30.9	29.3	31.9	30.3	30.4	28.3	29.9	30.2	32	27.1	27.7	31.6	29.5	29.2	30.7
32.1	31.3	33.5	32.1	32.1	29.9	32	31.9	32.9	29.1	29.9	33.5	31.2	30.8	32.2
28.4	26.6	29	27.3	28.6	25.8	27.2	27	30.2	25.3	25.7	28.2	26.6	26.2	28.2
26.1	22.9	25.4	27.3	27	24	26.4	26.5	27.4	24.2	26	25.7	25.2	24.5	26.3
26.1	22.4	24.9	27.3	26.7	22.4	27.4	26.4	27.6	23.9	25.5	25.9	26.2	24.1	26.4
26.1	22.5	24.9	26.3	27	23.7	26.2	26.5	26.7	24	24.5	25	25.4	24.9	26.3
26.7	23.7	26.4	27.8	27.8	24.4	26	26.8	28.5	24.2	25.3	27.1	26.6	24.8	27.4
25.6	21.9	24.5	26.6	25.3	22.9	24.2	24.5	25.7	23.2	23.5	24.5	23.9	23.6	25.9
26.4	21.6	24.7	25.9	27	22.7	26.2	25.7	26.2	23.9	24.1	24.6	24.8	24.7	26.1
27	23.8	26.2	27.8	26.2	23.5	26.6	26.2	27.4	24.6	24.5	26.2	25.5	23.5	25.5
24.9	21.7	22	24.6	23.7	20.7	25	24.5	25.1	22.2	22.6	22.9	23.2	22.8	23.5
26	23.5	26	27.2	27	23.8	25.7	25.9	27.1	23.6	24.8	26.3	25.7	24.2	26.9
27.2	24.6	26.4	28.4	27.5	24.5	26.6	26.8	27.8	24.1	25	26.4	26.3	24.7	26.8
26.2	23.2	26	27.1	26.5	22.9	25.9	25.8	27.1	23.9	24.8	25.6	24.6	22.9	24.8
27.6	24.8	26.8	28.5	28.1	25.5	28	27.6	28.3	25.5	25.9	26.1	26.5	24.3	27.3
26.7	24	27.4	26	27.6	25.7	27.2	28.1	28.3	25.5	28.3	30.4	27.3	26.5	28
23.5	18.7	22.7	22.5	23.2	21.1	23.6	24.5	25.1	23.3	23	26.7	23.4	22.6	26.7
25.6	23.4	25.8	25	27.6	24	25.4	27.4	27.9	25	27.1	28.4	26	24.9	27.5
24.8	20.4	23.5	23.7	25.7	22.4	24.5	25.8	27.7	24.6	24.1	28.3	25.5	23.4	26.9
29	25.4	28.2	27.6	29.3	27	27.9	28.5	31	26.4	28.8	30.7	28.3	26.8	29.4
28.7	25.3	28.6	27.1	29.3	26.8	26.9	29.1	30.4	25.5	28.7	30.7	28.2	26.5	29.4
25.4	22.6	25.9	25	26.3	24.2	25.9	27.9	28	25.4	26.7	30	27.3	25.1	28.8
23.2	20.2	23.3	22.9	24.2	21.3	23.4	25.6	25.9	23.5	25	27.3	24.6	23.4	26
26.9	23.5	27.3	26	27.8	25.5	26.9	28.6	28.7	25.3	28	29.6	26.6	25.9	28.4

23.5	20.1	22.6	22	23.3	21.3	22.8	24.9	24.9	22.5	23.8	25.5	23.5	21.8	25.4
25.5	20.9	25.8	25.2	25.7	23.6	24.6	26.5	27.4	24.7	24.8	27.7	25.3	24.6	27.9
28.7	24.8	28	27.6	28.6	26.9	27.5	28.9	29.9	25.4	28.9	30.7	27.7	27	29.1
27	24.2	27.1	26.8	28.3	26.4	27.3	28.3	28.8	25.5	28.2	29.6	27.4	26	28.6
26.4	22.5	24.4	24.9	26.3	22.1	24.4	25	26.4	23.5	24.5	26.5	24.4	24.8	26.2
28.9	25.2	28.5	28	29.2	27	27.6	28.5	30.6	26.7	29	30.5	27.5	27	29.7
26.8	23.1	27.5	26	27.6	25.5	26.3	27.6	28.7	25.2	26.9	29.3	26.6	25.3	27.7
23.6	20.8	23.6	23	24.9	22	23.9	25.1	26.4	23.7	25	28.7	25.5	23.9	26.6
22.6	19.2	23.2	21.9	23.4	19.4	24	24.1	25.1	23	23.6	26.9	24.3	22.1	25.2
29	25.5	28.2	27.6	28.9	26.8	28.3	29.1	30.6	26.5	28.9	30.7	28.1	27.2	29.9
28.2	24.5	27.3	26.8	28	26.5	27	28	29.3	25.6	27.6	29.7	27.4	25.5	28.7
28.4	24.4	27.6	26.7	28	25.3	27.2	28.4	30.5	25.6	27.7	29.6	27.2	25.8	28.3
27.1	24.7	28	26.4	28	26.2	27.7	29.6	29	26	28.7	30	28.1	26.6	28.5
26.1	22.8	26.4	25	27	24.4	26.4	27.2	27.5	24.4	27	28.1	25.9	24.7	26.9
22.7	18.6	23.2	21.8	23.5	20.2	22.5	24.2	25.4	21.7	22.2	25.3	22.9	20.4	24.8
20.5	19.1	19.3	20.8	22.3	18.7	20	23.1	23	21.2	22.1	23.8	21.5	20.8	22.1
23	20.3	22.2	22.8	24.4	23.2	23.4	24.5	25.4	23.9	25.2	26.5	24.9	23.4	26.4
25.3	21.4	23.2	24.3	24.6	24.2	25.2	26.2	26.7	25.4	27.5	28.3	27.3	25.3	27.8
24.9	21.4	22.5	23.6	25	24.5	24.8	25.8	26.1	27	27.2	27.9	26.1	24.6	26.4
24.3	20.7	22.3	23.3	24.2	24	24.6	25.4	26.5	25.3	26.4	27.8	26.1	24.6	26.6
23.5	20.6	22.2	23.1	24	22.6	24.7	24.8	26.3	24.4	25.3	27.1	25.4	24.3	27.2
24	20.7	22.4	23.6	24.6	24.1	24.3	25.7	26.2	25.2	26.1	26.9	25.6	24.4	26.5
25	20.9	22.5	23.6	25.5	24.9	25.1	26	27	25.8	27.7	27.7	26.5	24.7	27.6
28.5	24.8	26.8	27.5	30.5	28.1	28.3	29.4	31.3	29.2	30.7	30.8	28.9	25.4	30.5
27.3	23.8	25.6	26.7	28.7	27.1	26.8	28.1	30.3	27.2	29.8	29.7	28.1	27.3	30.3
25.6	22	24	24.8	25.6	24.8	24.8	26.9	27.8	26.5	27.3	28	27.3	25.7	27.7
26.6	23	24	24.5	26.7	25.5	26.8	26.9	27.5	27.2	27.2	29.3	27.9	26.2	27.9
25.2	21.1	22.7	23.8	24.9	24.4	24.2	25.7	26.2	25.9	26.5	27.6	25.9	25	26.5
24.1	20.4	22	22.4	23.9	22.9	24.5	24.2	25.7	24.5	25.5	26.9	25.3	23.4	26
26.4	22.5	24.1	25	26.3	26.1	26.1	26.7	27.8	26.9	28.4	28.3	26.8	25.1	27.7
25	22.1	22.2	23.2	24.7	24.9	25.1	26.4	26.6	25.5	27.3	27.1	26.1	24.5	26.9
26.3	22.4	24.1	25	26.3	25.8	25.8	27.2	28.6	26.9	27.6	28.6	27.4	25.7	27.4
27.4	23.5	25.8	27	28	27.3	27.5	29.2	30.4	28.9	30	30.9	29.8	27.5	30
27.4	22.9	24.1	25.6	26.8	26.5	26.3	28.3	28.3	28.3	28.9	28.7	27.6	25.9	26.6
27.4	23.1	26.1	25.5	28.3	25.8	27.3	27.5	29.7	27.2	29.5	30.2	28.1	26.4	28.9
26.6	23.4	25	25.5	28	26.3	25.9	27.2	29.4	26.9	29	29	27.3	26.5	28.5
26	23.2	25.3	26	27.3	25.3	25.4	27.2	28.3	26.6	27.9	28.1	26.6	25.4	28.3
26.7	22.5	23.1	25.3	25.5	26.2	26.1	27.2	27.1	26.4	27.5	27.9	26.3	25	27.1
25.8	21.5	23.9	24.6	26.8	25.2	25.5	26.6	28.4	25.5	27.8	28	26.3	25.3	27.7
26.4	22.4	23.2	24.6	26	26.2	27	26.9	28.8	28.3	29.7	27.9	26.7	24.9	27.9
28.2	23.6	26	25.9	28.2	26.8	27.3	27.9	29.8	26.9	29.5	30.7	28.4	26.7	29.2
27	23.4	25.2	25.8	27.7	26.2	26.1	27.1	28.8	26.6	28.7	28.7	26.8	25.9	28.5
28	23.5	25.7	25.2	28	26.5	26.9	27.7	29.4	26.8	29.2	30	27.6	26.4	28.9
27.8	23.8	26.3	26.5	28.4	26.6	27.1	28.2	30.2	27.3	29.7	30.3	28.1	27.1	29.4
27.1	22.6	24.8	25	27.5	25.4	26	26.5	28.3	25.9	28.4	28.7	26.8	25.8	28.1
27.7	23.1	25.3	26.8	27.5	26.6	27.2	28	29.4	27.3	29.2	29.4	27.7	26.4	28.9
27	22.7	24.8	25.5	27.7	26.3	26.2	27.6	29.3	27.2	28.6	30.3	28.3	27.1	28.1

29.2	22.6	25.7	26.5	28.7	26.9	28.2	28.5	29.9	28.5	29.3	31.4	28.8	26.6	28.8
31.3	29.1	28.5	29.2	30.2	28.9	31.1	29.4	31.5	28.8	30.5	31.6	31.4	27.7	31.1
31.6	30.3	31.4	30.5	30.3	30.6	31.1	29.8	32.8	29.1	31.5	32.3	31	29.1	31.1
32.4	30.6	31.1	30.8	33.9	28.7	31.1	28.3	32.4	29	31.4	32.4	31.1	28.9	33.2
30.2	29.2	29	29.4	29.2	27.5	29.7	30.6	31.2	27.6	30.1	31.1	29.4	28.9	31.7
32.2	30.2	31	30	31.5	29.6	32	30.2	34	30.3	31.7	32.6	31.9	28.9	33.5
34.3	30.2	31.5	30.6	31.7	30.1	31.4	31.8	32.3	31.1	32.5	32.1	32.5	28.8	32.5
33	30.4	31.2	29.9	30.7	30	32.3	30.8	32.5	30.3	31.3	31.7	31.5	28.6	33.6
31.7	29.4	31.8	30.2	30.8	30.3	32.2	31	31.8	29.3	32	30.8	31.7	28.5	31.4
32.4	30.9	33.4	30.7	31	30.2	31.4	30.8	32.9	30.6	31.1	32.1	31.7	29.9	31.3
32.9	31.2	32.2	30.4	31.8	30.1	33.4	30.8	33.8	30.2	32.1	32.5	31.6	29.1	33.5
33.2	31.3	34.6	31.1	30.7	31	32.6	32.1	33.4	31.5	32	32.4	32.1	30.3	31.7
32.7	27.9	28.7	28.4	29.9	28	29.8	28.8	31	28.7	30.5	31.2	31.9	28.2	30.9
31.4	29.9	30.8	30.4	31.9	30.2	31.9	30	32.9	29.2	31	32.2	31.3	28.8	32.8
29.8	27.5	27.8	28.3	29.9	28	29.5	28.6	31.8	27.8	30.6	30.5	29.6	27.2	31.4
30.9	28.2	28.1	29	30.6	28.9	31	29.8	32.2	28.9	30.9	32.4	32	27.7	31.3
34.2	30.4	31.7	31	31.9	30.4	31.9	32.6	31.9	30.8	32.8	31.8	32	28.8	32.8
34.5	31.6	32.8	32.3	33.1	31.4	32.5	33.3	32.7	31.6	33.7	31.9	32.2	30.5	32.3
34.3	30.5	31	31	32.2	30.5	33	31.6	33.2	30.7	32.3	33.3	34	28.5	32.8
34.5	31.8	34.5	30.8	31.7	31.8	32.9	32	34.2	31.6	32.9	32.6	32.9	30.9	32.3
34.5	30.7	31.6	31.9	32.3	29.4	33.1	32	33.3	30	31.7	33.1	34.2	29.1	33.8
30.5	28	28.1	28.4	30.5	29.1	31.2	29.6	31	29.2	30.5	32.4	31.6	27.3	30.5
32.8	29.9	30.2	30.5	31.2	29.9	32.2	30.6	33.2	30.2	31.8	32.8	32.9	28.5	32.2
30.5	27.6	28	28.6	30	28.4	30	29.5	32.4	28.7	30.7	31.6	30.8	27.8	31.5
33.3	30.1	30.7	30.6	32	30.1	32	31.3	32.9	30.4	31.9	32.4	33.6	28.2	31.5
33	29.3	29.5	30	31.5	30.1	31.8	30.6	32.4	29.8	31.4	32.6	32.6	28.9	31.9
32	30.1	30.2	30.4	31	29.8	31.8	29.9	33	30	31.7	32.1	31.5	29.1	32.1
31.6	31.5	31.6	30.6	33	32.4	32.4	33.3	32.3	35	31.8	32.6	31.1	32.2	32
32.5	32.5	34.1	30.8	34.1	33	32.2	33.8	33.3	34.9	33.2	33.1	31.8	32	32.5
33.6	33.9	33.6	31.9	35	34.5	33.9	35.2	34.8	37.2	33.7	33.9	33.2	33.8	34.1
33	33.1	31.3	31.6	34	33.1	33.1	34.2	33.4	35.5	33.5	33.3	32.3	32.2	34.8
34.1	32.6	34.6	30.6	33.4	32.1	33.1	34.1	33.3	34.7	33.3	33	31.9	32.6	32.4
33.1	32.6	33.5	30.9	33.2	31.6	32.5	33	32.7	34.5	32	32.3	31.8	31.5	31.7
34.9	34	36	32.2	34.9	34.1	34.4	34.9	34.9	36.7	35.1	34.3	33	33.7	33.4
32.7	31.5	34.2	30.1	32	31.3	32.6	32.8	33.4	33.1	32.7	32.7	31.5	31.1	31.3
33.3	32.1	34.7	31.1	33	32.7	33.5	33.6	33.7	33.8	33.5	33	31.8	31	32.8
31.6	31.4	31.4	30.1	31.9	31.5	31.1	32.4	31.6	33.7	31.3	32.3	30.3	32.3	31.8
31.5	29.1	32.6	28.8	31	30.5	30.7	31	31.3	31.2	31.1	31.8	29.3	30.8	30.9
32.3	32.5	32.3	30.8	32.9	32.3	32	33.4	32.4	35.1	31	31.6	31.5	31.4	32.2
35.2	35.3	35.6	33.6	36.5	35.8	35.6	36.6	35.8	38.2	35.5	35.3	33.3	33.9	35
33.4	32.8	33	31.9	34.3	33.5	33.3	34.2	33.3	35.4	32.6	32.8	32	32.8	33.2
33.5	33	33.2	32	33.7	32.3	33.2	34.3	33.4	35.5	33.2	33.3	32.8	32.3	33.9
33.1	34.1	33.5	32.1	34	33.7	33.8	35.8	33	35.8	33.6	33.9	32.6	32.5	34
33	33	33.6	31	33.4	32.9	33.4	33.8	33.4	36	33.2	33.5	32.2	31.8	32.8
31.9	31.8	32.2	30.3	32.2	31.6	32.7	33.8	32.1	34.3	32.9	32.7	31.6	31.8	32.8
32.2	32.2	32	30.8	33.1	32.9	33.2	33.2	32.4	34.6	31.9	32.2	31.1	31.1	32.2
33.3	32.1	34.3	30.9	32.7	33	33.5	34	33.1	33.8	32.8	32.7	31.8	31.7	33.4

32.8	33	33.7	31.8	34.1	33.7	33.1	34.8	32.3	35.9	32.6	33.4	31.5	31.8	32.4
33.5	32	34.3	30.7	33.1	32.5	33.1	33.4	33.5	34.1	33.7	33.5	32.6	31.1	32.4
34.8	34.4	35.2	33	35.1	34.7	35.4	35.5	35.1	37	35.1	35	34.2	33.7	34.4
34.1	32.7	35.1	31.3	33.8	32.5	33.6	35.4	34	34.8	33.8	34	32	32.4	33
34	32.8	35.5	32.2	33.8	33.4	34.2	33.9	33.8	34.5	34.2	33.9	32.7	32.3	33.4
30.6	31.6	31.3	30	31.6	31.8	31.4	32.9	31.2	33.6	30.9	31.3	30	31.2	31.1
33.5	32.6	34	32.2	35.8	34.3	33.6	35.3	34.3	36	34	33.8	32.6	33.4	33.2
33.9	31.6	34.8	30.5	32.8	32	32.8	32.3	32.8	34.1	32.6	32.2	30.9	30.9	31.9
33.6	31.9	34.6	30.8	32.8	32.2	33	34.4	33.1	33.9	33	33	31.8	31.2	33.4
33.1	31.9	33.9	30.5	32.8	32.3	33.8	33.9	33.4	34.7	33.2	33	32	31.3	31.1
34.1	32.8	35.2	31.5	33.6	32.5	34	35.2	34.6	34.8	33.6	33.8	33.5	32.4	33
33.2	32.2	34.1	30.9	32.9	31.4	32.6	33.9	33	33.6	32.8	32.7	31.8	30.1	32.1
29.9	24.7	20.8	29.5	26.4	30	26.4	29.8	27.9	30	27.5	29.9	32	29.3	30.2
27.3	23.2	18.9	26.5	24.8	27.2	24.5	27.2	26.6	27.7	25.5	28.9	30	26.5	27.9
29.8	27.1	22.5	33.4	29.1	31.1	28.4	33	28.3	30.6	30.8	30.8	32.6	30.2	29.7
29	24.2	21.8	29.1	27.6	29.8	28.3	32.2	31	32.9	31.3	33.7	33.8	30.1	31.8
29	24.9	21	30.4	27.1	29.5	26.2	29.5	27.7	29.8	28	29.7	32.5	29.8	30
26.8	21.5	18.8	26.9	23.2	26.2	24.8	27.7	24.6	29.2	27.2	27.2	29.6	27.3	27.3
26.8	22	17.9	26.3	23.5	26.9	23.3	26.9	25.3	27	24.6	27.9	29.5	26	27.3
26.6	21.9	18.1	25.6	23.6	26.7	23.5	26.8	25.5	27	25.4	27.4	29.4	26.5	27.8
27.7	23.4	19.4	30.7	24.3	28.5	25.9	30.3	26.4	28.4	27.3	29	31.4	27	27.8
27.2	21.8	19.2	26.9	24.1	27.5	23.7	28.1	25.9	29.2	28.7	29.5	31.4	27.4	29.7
27	22.8	19.5	26.8	24	27.2	24.4	27.8	26	27.7	26.5	28.9	30.3	25.2	27.6
29.2	23.9	20.3	27.6	26	28.1	25.2	27.3	27.1	28.9	27.1	28.9	30.6	27.6	29.2
26.9	21.7	17.9	25.2	24.3	25.9	22.9	26.4	25.8	26.5	25.1	27.2	29	25.2	27.3
29.5	26	21.8	31.7	28.4	31.4	27.9	30.9	28.1	29.9	28.1	30.8	32.6	30.4	30.6
28.8	24.1	22.4	29.6	27.2	29.4	28.1	31	28	30.1	28.7	31.1	32.5	28.1	31
29.5	24.1	21.2	29.4	26.3	29.6	27.2	30	28.3	31.3	29.4	31.2	32.9	29	31.2
27.4	24.5	19.5	30.2	25.7	28.7	25.5	28.9	26.1	28.6	27.3	29.4	32.2	28.5	28.4
28.3	22.6	20.6	28.1	27.3	28.6	26.9	30.2	28.4	30.7	29.3	31.7	32.7	29.1	31.3
31.3	26	22	30.3	28.3	30.3	27.4	30.3	29.2	31.5	30.3	30.9	33.6	30	31.8
27.6	22.3	19.4	26.9	26.5	27.2	23.6	27.1	26.1	28.3	26.5	28.3	30.2	26	28.2
25.9	20.7	19.6	25.5	24.2	26.4	23.2	27.5	27.4	27.9	27.5	29.5	30.2	26.3	28.7
29.9	23.8	21.9	30.5	28.6	30.2	26.9	32.1	31.2	33.1	31.5	34.1	33.9	29.9	32.1
27.9	22.1	18.2	26	24.4	26.4	23.4	27.4	26.8	28.9	26.6	27.9	30.2	26	27.1
30.7	24.2	20	28.6	27	29.4	25.3	28.7	28.4	30	28.6	29.7	32.2	28	30.3
26.2	20.6	17.7	27.2	23.8	25.5	23	26.1	24.5	27.7	25.2	27.8	29.5	25.6	27.6
25	20.5	17.9	24.2	22.5	25.4	23.2	26.2	24.6	26.1	24.1	27.1	28.8	23.5	25.1
29.6	24.1	22.3	29.8	26.6	28.9	27.9	31	27.7	30.9	31	30.6	32.3	28.9	30.6
29	23.8	20.4	29.4	25.4	29	26.7	29.1	27.1	30.9	29.1	30.4	32.8	28.8	30.4
30.7	24.8	22.4	28.9	28.1	30.7	28.1	30	28.8	31.3	29.9	31.8	33.1	29.5	31.4
26.4	21.5	19.4	28.1	24.2	26.6	24.5	27.5	24.8	28.7	26.8	27.8	28.9	26.3	27.5
29.5	24.4	22.8	30.5	27.6	30.6	28.3	31.7	27.8	31.8	30.9	32.1	33.3	28.8	30.7
28	22.3	18.9	25.5	25.3	26.9	24.4	27.9	26.9	28.7	26.4	28.1	29.1	25.8	29.3
30.2	25.5	23.2	31	30.3	30.8	29.2	32.8	31.7	34	30.6	34	33.8	30.1	32.2
30.1	24.2	22.3	30.5	29.5	30.2	29.2	31.9	30.5	31.9	30	31.9	32.6	29.8	30.9
28	23.1	22.5	29.6	26.3	29	27.7	30.6	27	30.1	29.8	30.3	32	27.6	30.2

27.9	22.9	20.1	29.2	25.3	27.7	25.9	27.7	26	28.9	26.8	28.6	30.3	26.8	28.7
29.8	23.2	21.8	28.2	28.5	27.9	26.7	30.3	28.8	30.6	28.1	32.2	31.3	29.1	31.1
28	23.1	22.1	27.2	26.6	27	27.2	28.3	26.2	28.3	28	29.7	30.8	26.1	27.7
28.7	22.5	21.5	30	24.3	28.1	27.5	30.2	26.5	30.6	30.3	29.8	31.9	28	30.3
31.5	25	23	30.2	30.2	30	28.8	31.6	30	32.7	30.9	33.3	33	30.6	31.3
28.3	23.3	21.8	28.2	28.7	29	26.7	30.2	28.8	30.7	29.1	32.5	31.6	29.3	31.2
30.5	23.3	21.9	29	29.4	28.3	28.4	31.2	28.5	31.5	29.6	32.9	31.5	29.7	31.5
31.9	26.4	24.8	30.5	30.8	30.2	28.9	31.3	30.5	32.1	30.6	33.9	32	30.2	31.8
29.2	24.8	23.2	30.2	29.4	30.1	28.4	31.1	28.4	30.6	29.2	32.3	31.9	28.6	29.4
26.4	26	29	25.2	27.3	25.1	26.5	27.1	27.6	25.3	24.7	26.6	26.2	26.2	27.1
30.8	31.1	31.8	31.2	30.5	31.2	30.7	29.8	30.2	28.2	28.1	30.2	29.1	31.1	31.2
31.8	33.2	35.7	30.6	32.7	31.7	32.4	33.1	32.7	31.5	30.4	34.2	30.8	31.5	32.4
30.4	30.2	31.5	30.1	29.9	29.6	29.8	30.7	31.5	29.9	29.6	31	30	31	31.1
31.4	31.6	33.7	29.3	30.9	30.7	30.9	31.6	31.1	30	29.1	32.2	30	30	30
32.8	32.2	34	32.2	32	30.9	32.4	32.2	32.2	29.4	30.1	33	30.6	31.9	33.6
34.7	35.1	36.5	33.7	34.7	32.3	34.3	34.4	33.8	31.1	31.2	34.9	32.1	33.3	34.8
32.6	32.1	33.9	32	32.1	31.1	31.9	32.3	32.1	29.5	29.5	33	30.4	31.7	33.2
30.6	31.4	33.1	29.8	31	30.7	31.3	31.4	31.2	29.4	28.7	31.5	29.4	30.5	30.8
32.7	34	35.5	31.6	32.3	31.5	32.5	33.5	32.3	30.4	30.1	34.3	31.4	32.6	33.6
31.4	31.7	32.1	31.3	32.6	30	30.9	31.2	30.6	28.7	29	31.3	30.8	31.2	31.3
24.6	25	28.2	23	25.4	24.6	25.2	25.9	25.3	25	23.9	25.2	23.4	24.7	24.9
32	32	33.3	30.4	32	31.8	31.6	32.3	31.6	28.7	29.6	32.3	30.1	30.7	31.8
32.9	32.4	34.1	31.9	32.2	31.3	32.5	32.5	33	29.8	30.7	35	31.4	30.5	33.8
33	33.2	34.5	31.1	32.8	32	32.5	33.1	33.4	31.7	31.4	32.7	32.1	33.4	33.7
31.1	31.5	34.5	29.6	31.3	31.1	31.9	31.6	32.4	30.4	28.2	32	29.8	30.1	30.5
30.8	31.1	32.7	30.1	30.9	30.4	31	31.5	31.4	29	28.6	30.9	29.3	29.9	30.4
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33.6	33.3	35	32.5	32.4	31.1	32.5	32.8	32.4	30.3	30	32.7	31	31.8	32.9
31.5	32.7	34.8	30.2	31.3	31.3	31.6	33.5	31.2	30.2	28.8	32.7	29.9	31.1	32.3
27.5	27.2	30	26	28.5	27	27.7	27.7	28	26.6	26.6	28.1	26.4	26.6	27.6
33.1	33.9	34.1	32.8	32.3	32.1	32.4	32.3	32.2	30.6	30.3	32.7	31.1	31.6	32.6
31.2	32.4	34.9	30.4	31.8	31.4	31.8	32.7	32.6	30	29.1	32.3	30	30.6	31.2
33	33.9	35.3	31.9	32.8	31.6	32.7	32.9	32.3	30.5	30.1	33.6	31.1	32.1	32.7
32.1	33.6	35	31.8	32.5	31	32.6	32.8	32.5	29.7	29.3	33.1	31.4	32.2	33.2
29.8	29.3	32.6	28.3	30.1	28.4	29.9	30.2	29.4	29	28.2	30.2	28.3	28.7	28.7
28.5	29	31.5	26.5	29.1	27.7	28.4	28.9	29.1	28	26.5	28.9	27.2	27.6	28.3
30.7	30.8	33.9	29.6	31.6	30.2	30.7	31.2	30.8	29.4	27.8	31.3	28.8	29.3	29.8
34.1	34.3	36.2	33.2	33.4	31.4	32.7	32.9	32.7	30.4	30	33	31.2	32.4	33.8
28	21.2	21.1	26.2	26.2	25.2	25.7	28.9	28.1	27.7	27.7	30.6	30.3	28.1	29.4
28.8	23.2	21.9	26.9	27.1	28.3	27.8	31	29.9	30.1	30.2	32.5	31.9	28.6	30.3
28.2	21.5	21.9	27.1	26.9	27.3	27.1	30.4	28	29.4	28.8	31.9	31.5	28.3	30.6
23.6	19.5	20.6	22	23.6	23	23.6	25	24.7	25	25.2	26.3	25.2	23.3	24.6
29	23	22	26.7	26.8	26.5	26.4	28.9	28.4	28.7	28	31.3	30.4	28	29.1
25.6	21.9	20.6	24.2	25.2	24.4	25.8	26.2	25.7	26.2	25.8	28.4	27	25	25.5
29.4	22.9	22	24.7	26.5	27.1	26.7	29.5	29.2	28.9	28.4	30.5	31	28.8	28.9
26.6	22.4	22.9	24.2	25.7	25.4	25.4	27.2	27.5	26.6	27.4	28.1	26.3	25	27.3
27.8	22.7	21.9	26.2	26.5	25.8	26.3	27.8	27.7	27.9	27.5	29.6	28.5	26.5	27.3

28.1	22.7	21.5	24.4	25	24.7	26.1	25.6	27.4	25.7	27.1	28.8	28.3	26.4	27.9
27	22.8	21.6	25.2	26	24.6	25.9	26.9	26.4	27	27.5	29.3	28.3	25.8	26.4
26.1	21	22.4	24	25.5	25.1	25.2	27.4	26.4	26.7	26.2	27.9	26.4	24.9	26.5
30.3	25.3	22.5	28.9	27.6	27.1	27.1	29.7	28.1	29.8	28.5	30.9	30.2	28	29.1
27.2	22.9	24.2	25.7	27.6	26.1	26.5	26.8	28.1	27.9	28.2	28.8	26.5	25.8	27.5
29.1	23	23.1	26.7	26.8	26.4	28.1	29.5	28	27.7	28.2	30.7	29.4	27.5	28.7
26.6	23	22.1	25.8	26.7	26.1	27.4	27.8	25.7	27.9	27.2	28.8	29	25.7	27.3
30.5	25.4	26.1	28.7	29.1	29.1	28.3	30.8	29.2	29.6	30.3	32	30.9	29.3	31.8
27.8	23.7	24.9	25.9	27.2	27.4	27	28.6	29.1	27.9	29.2	29.6	28.2	26.9	28.9
27.4	23.1	23.5	25.4	26.6	26.3	27	28.8	28.6	27.5	28.8	29.6	28	26.6	28.9
27.8	23.7	23.8	25.4	26.3	27.5	27.7	29.2	28.8	28	28.7	28.9	27.9	26.2	28.4
27.9	22.5	22.8	25.5	26.1	26.1	27.2	29	28.1	27.7	28.1	29.5	28.3	27	28.6
27.8	23.6	23.2	25.5	26.4	25.5	26.7	28.8	28.5	27.8	28.5	28.6	27.6	25.8	27.9
26.8	22.5	22.8	25.1	25.9	26.5	26.2	28.5	27.8	27.1	28	28.4	27.7	25.4	27.6
27.1	22.8	23.4	25.5	26.4	26.5	26.3	27.8	27.3	27.1	28.9	29.6	28.3	26.3	28.1
30.7	28.3	28.8	30.5	31.2	31	32	31.9	32.6	34.2	33.8	34.8	35.2	30.9	34.3
30.7	24.4	26.3	27.1	31.5	27.4	29.6	29.6	31.2	30.7	29.7	33.8	31	28.6	30.7
30	26.6	28.1	29.1	30.7	29.6	29.8	29.2	30.5	31.9	31.5	32.7	33.2	28.4	31.7
31.4	25.9	27.1	28.4	31.8	29.8	31.3	31.4	31.2	31.1	31.6	34.4	32.1	29.1	32
31.3	28.7	29.6	31.8	32.1	32	32.5	32.2	33.3	34	33.9	34.5	35.4	30.8	34.3
30.9	26.7	27.8	31	30.2	30.1	31.2	30.8	31.9	32.9	32.5	33.8	32.5	29.7	30.9
30.5	27.2	27.3	30.2	30.7	30	31.3	31.1	31.4	32.3	32.5	33.3	33.6	30.3	32.7
33.6	28.9	29.1	29.9	33.5	30.9	33	31.8	34.6	32.8	34.2	34.8	35.5	29.7	34.2
29.1	24.8	26.5	28.1	28.8	29.5	28.9	29.5	29.2	31.1	31	32	32.4	27.8	30.8
32.1	29.2	29	30.5	31.7	30.3	32.2	32.4	32.9	33	33.8	34.4	34.4	30.3	33.7
31.7	25.9	27.6	29	31.6	30.5	31	30.8	31.1	33.5	32.2	34.2	33.6	30.4	32.4
33	29.5	29.3	30.6	32.7	30.2	33.2	32.2	33.4	33.6	34.8	35.4	35.4	30.9	34.1
32.2	27.2	27.8	28.5	32.2	28.8	31.8	30.5	32.4	31.5	32.8	34.6	33.1	29.8	32.6
30.5	27.4	28.3	28.8	32.8	28.8	31.1	30.6	32.6	29.7	31.9	34.4	32.3	28.2	32.9
31.8	26.8	28.4	28.6	31.8	28.6	32.2	30.7	31.9	30.2	32.5	33.2	32.6	28.2	31.7
30.2	24.8	26.6	27	30.6	27.5	29.8	28.9	30.5	29.8	30.9	32.7	30.9	28.5	30.9
31.4	27.9	28.9	29.2	33.2	29.6	31.9	31.2	32.8	30.3	32.2	34.1	31.9	29.1	33
31.3	26	26.5	27.3	30.2	27.9	30.5	29.9	30.5	30.4	31.1	33.8	31.1	27.9	31.3
31.9	27.2	27.9	28.7	31.6	28.4	31	30	31.6	30.2	32.7	33.3	32.9	28.9	32
31	26.5	27.2	28.6	31.1	28.6	30.8	31.5	31.2	32.8	33.2	34.4	33	29.8	32.5
33.5	28.4	29	29.3	32.4	29.8	32.3	30.8	32.6	32	33.6	34.3	34.4	29.8	32.9
32	28	29.1	29.2	32.7	29.5	31	30	32.3	30.2	31.2	34	33	29.6	32.4
29.5	24	25.8	25.9	30.4	26.1	28	27.9	29.4	27.2	28.9	31.2	29.4	27	30.5
31.6	27.1	27.9	29.1	32.2	29	31.1	30.5	31.1	31.1	32.3	33.9	31.5	28.4	31.4
33.2	28.9	29.1	29.9	32.5	30.3	32	30.9	32.7	31.2	32.6	33.9	33.8	29.1	32.1
33.7	28.8	29.2	29.7	33.9	29.9	31.5	30.4	32.9	31.8	32.9	34.6	33.7	29.2	32.9
31.4	27.1	28.5	28.8	30.8	28.4	31.2	30.7	31.9	31.3	33	33.8	32.5	29.4	33.2
33	29.1	29.6	31.1	33.7	30.5	34	33.5	33.7	33	35.9	36.7	35.9	31.1	34.8
32.3	26.8	28.8	29.2	32.5	29	31.7	30.5	31.7	30.4	31.7	33	32.2	27.7	31
30.8	26.6	27.2	27.6	31.8	27.5	29.9	30	31.5	29.4	30.8	32.7	31.4	27.6	32
33	28	29.3	29.6	33.6	29.9	34.2	31.7	33.5	31.3	33	34.2	33.5	29.3	33.3
35.4	30.1	30	31.4	34.6	30.8	33.5	32.5	35.2	33.8	35.2	35.7	35.9	30.6	35.3

30.5	25	27.1	27.1	31.5	26.9	29.5	29.3	30.5	29.6	30.9	32.6	31	27.3	30.6
30.1	25.9	26.5	27.5	31.3	28	29.5	29.4	30.5	29.2	29.9	32.2	30.5	27.1	29.8
29.6	26.6	27.3	27.7	31.1	27.4	29.4	29.9	32	29.1	30.9	33	32.5	27.9	31.6
31.2	26.1	28.2	28.2	31.4	27.4	30.4	29.5	31.8	29.9	31.5	32.4	32	28.4	31.4
32.8	28.5	28.9	30	32.1	30	32.5	31.2	33.2	31.1	32.8	33.6	33.5	29	33.9
31.6	29.1	29.1	29.9	31.8	29.5	31.8	29.6	33.3	28.5	31.3	34.4	33	27.6	31.8
29.8	24.4	25.6	26.2	30	27	29	28.2	29.5	28.1	29	32.5	29.2	26.3	29.8
30.9	25.3	27.4	27.8	31.4	27.8	29.7	29.9	30.7	29	29.8	32.7	30.2	27.7	30.6
32.2	27.8	28.7	28.3	31.8	28	31.1	29.6	31.8	29.5	31.7	33.3	32.7	27.6	32.4
32.1	28.4	29.4	30.1	33	29.8	31.9	31.1	32.9	30.4	32.2	34.4	33	28.7	33
32	27.4	31	29.7	32.9	29.3	31.6	31.2	33.2	30.5	32.5	34.7	33.4	29.6	33.2
31.9	28.3	28.8	29.4	32.4	29.5	31.7	30.4	32.6	30.4	32.3	34.4	33.1	28.7	33.2
31.2	29	29.1	29.1	32	29.4	31.2	30.6	33.1	30.4	32	33.9	32.8	28.4	31.9
31.4	26	27.8	27.1	30.4	27.5	29.8	29.4	30.3	30	30.7	32.8	30.9	27.6	30.7
26.8	24.6	28	28	28.6	26.1	27.3	27.2	28.8	25.4	26.3	28	28	26	28.5
26.9	24.9	28.1	29.1	28.3	26.4	27.4	27.1	28.8	24.5	25.8	28.1	27.8	25.1	27.8
23.1	21.3	23.6	25.6	25	22.6	22.6	23.3	23.9	21.4	21.8	23.7	23.8	23.5	25.4
26.5	24.4	26.3	27.6	27.8	25.5	25.9	26.1	27.7	24.7	25.7	26.2	26.6	25.6	27.2
28.4	25.4	29.2	29.4	29.9	26.9	27.9	28.6	30.8	27	26.6	29.4	28.7	26.6	29.3
28.8	27.4	30	30	29.7	26.4	29.5	29.3	32.3	26.1	27.6	31.2	30.2	28	29.4
29.7	28.3	30.4	31	30.8	27.7	30.1	30.1	32.3	27.5	28.8	31	29.9	28.4	30.4
29.6	27.2	30.4	29.5	28.8	26.9	29.2	28.6	31.7	26.2	27.3	29.9	28.7	27.1	29.7
29	27.5	30.2	30.2	29.7	27	30.1	28.8	31.8	26.4	27.9	30.5	28.8	27.5	29.6
29.6	27.7	31.1	30.8	30.4	27.3	30.2	30.1	33.1	27.4	28.5	31.4	29.7	28.1	30.1
28.2	26.3	28.5	28.1	27.6	25.4	27.8	27.6	29.3	25.1	26.8	29.1	28.3	25.7	28
29.9	27.9	30.8	30.5	30.7	26.9	30.3	29.8	33.4	27.2	27.2	30.8	29.3	28	30
30	28.2	31.8	31.4	30.4	27.9	30.9	30.5	33.3	27.4	28.8	31.4	29.8	28.5	30.7
29.1	28.8	30.8	30.2	30	27	29.6	29.6	31.7	26.7	27.6	30.7	28.9	27.8	29.3
26.7	26.1	28.2	26.3	27.2	25.1	26.7	27.5	28.7	24.6	25.1	27.9	26.2	26	26.8
28.7	26.1	29.7	29.2	29	26.1	28	27.7	31.3	25.3	26.3	29.2	27.7	25.9	28.7
30.1	29	31.6	31.2	31	28	31.1	29.9	32.4	28.1	28.5	31.7	29.5	29	30.6
32.5	32.8	35.1	35.8	33	34.2	34	33.6	32.9	34.9	34.5	34.3	37.3	33.7	36.4
31.9	34.8	35.6	36.6	36.4	34.6	35.6	36.6	34.1	36.5	37.5	34.6	36.7	33.8	36.5
31.8	30.7	32.4	33.7	31.9	33.7	31.8	32.7	30.9	33.6	32.9	34.3	36.7	32.9	34.7
30.3	33	35	35.8	34.9	34.3	34.3	33.3	32.7	34.5	33.9	33.3	36.9	34.1	35.4
31.3	34.6	34.6	35.7	34.7	32.4	35	36.5	34.7	36.1	37.2	34.2	36.8	33	35.1
21.5	21.2	23.9	24.4	22.5	23.6	22.6	22.2	22	23.6	23.3	23.1	26.3	22.6	25.2
30.1	33.3	33.8	34.7	34	33.5	33.2	33.3	32	34	34.4	32.6	35	34.1	34.8
33	35.2	35.9	37.2	34.6	34.8	34.7	34.6	32.3	35	35.1	34.5	36.9	35.2	36.5
21.5	26	26.2	27	26.6	25	25.5	24.1	22.9	25.8	26.7	24.8	27.1	25.8	27.7
32	34.7	33.6	36.7	35.8	35	35.1	35.9	33.2	35.6	35.7	34.7	37.3	35.2	37.1
27.6	31.4	31	33.5	32.6	31.1	31.4	31.3	29.9	32.4	33	31.4	34.3	32.3	33.6
27.2	27.5	29.9	30.8	30	29.4	29.1	28.9	26.7	30.5	28.6	29.8	33.8	29.6	31.8
30.6	33.5	34	34.5	34.2	32.5	34.3	35.9	34	35.2	35.5	34.4	36.9	33	35.6
29.1	30	32	32.8	32.2	31.9	32.1	30.4	28.5	31.6	30.8	31.3	34.4	31.8	33.7
32.4	35.5	36.1	36.4	36.2	35.3	36.3	36.1	33.1	35.3	35	34.7	37.1	35.5	37.2
31.7	33.9	34.2	34.7	35.2	33.1	34.7	35.7	34.3	34.9	37.4	34.3	36.3	32	34.5

29.9	32.8	32.8	33.9	33.5	32.2	32.5	34	31.9	34	34.7	32.1	34.6	31.7	33.7
29.5	27	29.3	31.3	29.7	30.8	29	29.8	28.1	31.8	30.3	32.2	33.5	30.4	32.4
26.8	28.2	30.5	31	30.2	29.1	29.1	29.2	28.3	30.6	31.2	30	33.7	30.2	32.2
28.4	29.7	31.2	30.5	29.8	29.2	30.8	30.8	29.4	31.4	31.5	30.4	33.2	30.2	32.6
26.6	28	29.2	29.7	28.4	27.3	28.7	28.4	28.2	29.5	30	28.3	32.2	28.5	30.3
31.8	35.4	34.9	36.7	36	34.7	35.5	35.6	33.8	36	36.2	34.9	36.8	35.2	36.7
29.1	32.2	33.3	33.5	33.5	32.4	32.6	32.9	32	33	33.9	32.3	35.4	31.6	34
29.8	29.8	30.9	31.6	31.2	30.4	32.2	32.9	30.4	32.8	34.8	32.5	34.8	29.2	33.7
32.5	35.9	36.3	37.1	37.1	35.9	36.4	38.4	35.5	37.3	38.3	35.6	36.5	34.5	36.4
25.6	24.2	26.9	27.9	25.5	25.2	26.3	26.1	24.1	28	27.3	29.3	29.8	26.9	28.9
25.8	26.3	29.1	29	27.3	27.3	28.7	28.5	27	28.5	28.9	28.8	31.5	26.6	29.4
31.5	35.8	35.1	36.4	35.8	33.6	34.8	35.2	33.4	35.6	36.3	34.5	36.9	35	37
29.2	29.4	31.6	32	30.9	31	30.7	31.1	29.3	32	31.4	30.3	34.8	31	33.5
33.9	31.3	30.6	35.3	31.4	35	31.1	34	32.7	33.3	33.4	36.1	36.4	34.8	36.4
33.7	30.8	27.1	33.3	30.4	32.7	29.3	32.5	32.1	33.7	32.5	35.1	35.6	33.5	34.5
34.1	31.8	29.5	34.9	31.1	34.6	31.5	33.4	33.1	32.8	32.7	34.9	36	34.1	35.2
34.1	31.4	30	34.6	30.6	33.9	31.3	33.5	32.6	32.4	32.2	35	35.6	34.4	35.5
35.3	32.7	29.5	36	33.3	36.1	32.4	35.6	34.5	34.8	34	36.9	37.4	35.3	36.8
35.3	32.3	33	37	33.8	36.5	32.9	35.5	34.1	35.1	34.9	37.2	37	35.2	38
30.4	27.8	28.4	32.3	29.4	31.8	28.8	30.8	29.2	30.8	30.2	33.2	33	30	32.8
37.3	35.3	32	38	35.3	37.7	34.5	37.2	37.1	37.2	36.5	39.4	39.3	37.6	38.9
30.6	27.8	24.7	30.9	28.2	30.7	27	30.6	30	31.1	29.7	32.4	32.8	30.4	31.9
33.5	30.5	27.3	34.2	31.4	34.2	30.2	34.1	32.5	32.6	31.3	35	34.7	32.5	34.7
28.9	26.5	25.4	30.3	27.9	29.2	26.9	28.9	28.4	29.3	28.7	30.9	32.1	28.8	30.6
35.8	34.7	34.7	37.3	36.8	38.1	35.4	37.6	34.5	37.3	36.1	38.1	38.2	36.5	37.9
38.1	36.8	36	39.5	38.8	39.5	37.4	39.3	36.3	39	37.9	40	40.1	38.8	40.6
28	25	29.1	27.5	27.2	25.1	27.7	26.5	30	24.4	26.9	28.4	27.9	26.2	28.6
22.5	19.2	22.3	23	23.6	20.9	21.5	21.3	23.6	20.2	20.6	23.5	22.1	22	23.5
29.7	27.1	30.3	29.2	29.8	26.7	28.8	28.4	31.4	26	27.1	29.7	28.3	26.8	29.2
27.6	25.2	28.6	28.2	27	24.2	27.8	26.7	28.8	24.7	24.3	27.9	26.3	24.2	26.6
28.8	26.2	29.3	28.6	28.9	25.9	28.6	27.4	30.7	25.2	26.7	29.9	28.4	26.6	29.5
28.8	25.1	29	28.7	28.9	26.5	27	26.9	30.4	25.4	27.3	29.1	27.2	25.9	29.5
24.7	21.6	25.3	25.5	26.1	23.2	24	23.9	26.9	22.5	23.5	26	24.1	23.6	26.4
27	23.8	27.2	27.5	28	25.1	25.6	25.6	29	24	25.2	27.6	25.9	24.7	28
29.7	27.1	31.4	30.1	29.2	26.5	29	28.8	32.3	26.1	27.2	30.5	28.3	27.2	29.4
29.1	26.1	29.8	28.9	28.7	26.2	28.4	28.4	32.1	25.2	27.4	30.2	28.4	27.2	29.3
28.6	25.3	28.9	28.9	29.3	26.8	28.5	27.6	30.2	25.4	27	29.6	28.5	27	29
28	24	28.1	28.4	28.5	25.6	26.3	26.3	29.6	25.2	26.1	28.4	26.7	25.4	28.8
28.6	25.7	29.5	28.6	27.8	25.7	27.7	27.8	29.3	24.6	25.5	29.2	27.6	26.5	28.3
28.7	24.9	28.9	28.1	28.7	26.2	26.6	26.6	29.7	24.9	26.4	28.5	26.5	25.5	29.1
27.6	23.7	27.8	27	28	25.2	26.3	26.5	29.4	25.9	25.9	28.1	26	25.2	29.2
29.9	26.9	30.1	30.4	31.5	27.9	29.9	29.1	31.3	27	28.4	31.3	29.8	28.3	30.1
26.6	22.9	25.9	26.5	27.6	23.5	25.3	24.6	27.8	23	24.1	26.6	25.2	23.8	27.4
25.4	22.2	25.6	26.2	26.6	23.5	25.5	24.8	26.8	22.8	23.6	26.7	25.3	24.8	27.4
28	23.8	26.9	26.7	27.8	24.9	26.4	25.8	28.3	24.8	25.4	27.5	26.9	25.7	27.4
29.5	27	30.7	30	29.9	27	29.7	29.2	31.7	25.9	26.2	30.1	28.6	27.7	29.5
28.8	25.5	29.1	28.9	29.9	26.7	27.6	27	30.6	25.5	27.4	29.7	28.1	26.4	30.2

26.9	21.7	26.3	26.1	27	25.9	25.2	26.8	29.7	23.8	25.4	28.4	25.4	24.1	28.2
26.8	23.4	26.4	26.7	26.8	23.3	25.8	24.9	27.5	23.7	24.8	27.4	25.9	24.5	27.6
29.6	26.4	31.4	29	28.9	25.8	29.4	26.6	30.2	24.8	26.3	29.1	27.5	26.7	28.3
28.7	26.5	29.8	29	29.3	26.2	28.3	28.4	31	25.8	26.8	29.2	28	26.7	28.7
25.4	22.1	25.9	26.5	27.4	24.3	24.9	25.1	27.1	23.5	24.4	26.5	25.3	24.7	27
27.3	22.9	26.4	26.7	27.3	24.9	25.8	25.2	26.9	24.3	24.9	26.1	26.1	25.3	27.4
26.9	23.2	26.5	27.5	28.2	24.7	25.6	25.9	28.5	23.8	25.1	27.7	26.4	25.1	28.3
27.4	22.7	27	27.3	27.3	25.2	25.4	25.5	28.5	24	25.2	27.5	25.8	24.4	27.6
25.2	22.1	24.7	26.9	26.3	22.9	24.7	24.5	26.5	22.4	23.7	26.6	25	24	26.8
26.5	22.1	26.8	26.8	26.4	25	25.2	25.3	28.1	23.6	25	27.8	22.6	24.2	27.9
27.1	24.8	28	27.6	28.5	25.1	26.9	25.5	28.5	24.5	25.6	28.2	26.3	25.5	28.2
27.3	24.7	27.5	28.9	28.5	26	26.9	26.1	28.5	24	26.1	28.5	26.5	26.3	28.9
26.9	23.3	27.1	28.1	27.7	25.2	26.3	25.8	28	24.8	26.1	27.7	26.7	26.1	28.1
26.4	23	26.5	27.6	28	24.9	25.9	25.6	27.5	23.6	25.4	27	25.5	25.4	28.1
29.7	27.2	30.8	29.8	30.2	26.9	29.2	28.8	31.9	26.1	27.5	30.5	28.9	27.4	29.5
27.3	23.9	27.2	26.2	27	24.7	24.9	25.3	27.8	23.9	25	27.6	25	23.8	27.4
29	24.9	30.1	28.9	29.2	27.6	26.5	27.1	30.1	25.6	27.2	29.2	27.2	26.4	29.5
28.2	24.9	28.4	28.9	29.5	26.5	27.5	27.1	29.5	25.3	26.7	29.2	27.9	27.2	29.6
26.5	22.9	27	26.3	26.7	25.2	24.6	25.5	28.2	24.1	25.3	27.5	25.1	24.4	28.4
27.8	23.1	27.9	27.7	27.7	25.8	26.4	26.6	29.7	24.8	26.5	28.8	26.6	25.2	29.1
23.5	20.8	23.7	24	25.4	22.2	23.4	23.1	25.1	21.5	22.4	24.8	23.8	22.9	25.2
28	23.5	27.9	27.7	28	23.2	26.1	26.4	27.9	25	25.4	26.9	25.5	25	28.2
27.7	23.5	28.2	28.2	27.7	25.8	25.2	25.7	28.7	24.7	26.2	28.1	25.6	24.6	29.2
26.6	24	28	27.5	27.9	25.3	26.5	25.6	28.1	23.6	25	28.2	26.6	24.5	27.7
26.4	22.5	25.9	27.1	26.4	24.8	25.3	25.2	26.9	23.8	25.2	26.8	25.4	24.9	27.7
25.8	23.2	25.7	26.4	27	23.5	25.2	24.5	27.4	22.9	24	26.5	25.1	23.4	26.3
27.3	23.8	28.1	26.4	28.2	25.4	26.5	26.2	28.1	25	25.7	27.4	26.2	26.3	28.1
27.6	23.2	27.2	27.5	27.6	25.1	26	25.9	28.7	24.4	25.9	27.9	26	24.8	28.1
23.8	20.9	23.7	24.5	24.9	23.5	23.1	22.9	25.2	20.4	22.2	25.4	23.2	22.1	25.1
28.7	25.8	28.7	28.7	30	27.5	28.4	27.9	30	25.7	27	29.2	28.5	26.7	29
26.7	23	25.8	26.8	26.8	24.2	24.8	25.6	26.4	23.5	24.4	26	25.5	25.4	26.9
28.4	23.9	28.3	28.9	28.2	25.9	26.2	26.4	29.9	25	26.5	29.4	27.1	24.8	28.6
26	22.7	25.5	25.8	26.7	23.7	24.6	24.5	27	22.7	23.9	26.3	25.1	24.1	25.9
27.7	23.8	27.5	26.9	27.2	25.6	25.2	25.9	28.6	24.2	25.6	27.9	25.4	24.7	28.1
27.6	23.4	27.9	27.5	27.3	25.7	25.5	25.8	29.2	24.5	26.2	28.2	26.1	24.6	28.3
27.1	22.2	26.9	27.1	27.4	25.1	25.9	26.1	28.3	24.4	25.1	27.5	26	24.9	28.8
28.5	26.1	28.3	27	27.7	25.1	26.8	26.5	30	24.9	27.6	30.2	26.4	25.9	30.1
28.3	25.6	28.3	27.5	27.9	26.5	27.1	27.2	30.5	26.1	27.6	29.7	26.9	26.1	29.6
28.9	26.2	28	26.9	27.8	25.8	26.9	26.6	30.1	25.3	26.4	29.5	26.6	26.1	29.4
28.2	25	28.2	26.9	27.2	25.9	26.4	26.6	29.1	24.8	27.3	28.9	25.7	25.2	28.6
29.3	25.4	28.1	27	28.2	26.6	27.1	26.5	29.8	26.2	28.6	29.9	26.4	26.2	28.2
30.4	26.7	29.9	28.5	28.5	27.5	28.4	28.5	31.2	26.2	28.1	30.5	27.6	27	29.6
30.3	28.6	31.2	29.3	31.2	29	30.5	29.8	34.3	28.8	30	31.7	30.1	29.6	31.3
30	26.8	28.8	27.7	28.8	26.9	28.6	27.6	31.1	25.9	27.2	29.9	27.5	26.8	29.6
29.5	27.9	30.3	28.7	29.9	27.9	29.1	28.4	32.1	27.5	28.3	31	28.2	27.6	29.4
30.2	27.1	29.2	28.2	28.6	27	28.1	28.1	30.8	27.1	28.9	30.4	27.8	26.9	30
30.3	27.4	28.8	28.9	29.2	28.5	28.9	28.9	31.6	28.1	29.2	31.1	27.9	27.2	30.1

31.2	29.3	31.6	30	31.2	28.9	30.1	29.7	34.3	28.3	29.2	31.4	29.8	29.3	31.2
29.7	26.7	29.5	28.3	29	27.3	27.2	27.7	30.6	25.8	28.5	30.3	27.3	26.7	29.3
30.5	27	29.6	29.1	29.3	28.2	27.8	28.7	31.1	26.9	29.8	31.4	27.9	27.2	30.1
29.7	26.2	29.2	27.9	28.5	26.8	27.4	27.5	30.3	26	28.2	30	26.9	26.6	29.9
30.3	26.8	29.7	28.3	28.8	26.9	28.3	28.1	31.6	26.8	27.8	29.6	27.6	26.9	30.2
30.2	27.8	29.6	30	28.8	28.3	28.3	29	31.4	27.6	28.6	31.2	28.4	28.8	29.6
30.6	27.8	29.6	29	29.8	28.3	28.3	29.2	31.6	27.1	28.9	31.5	27.5	27.1	29.6
31	28.5	30	29.7	30	28.6	29	29.2	32.1	27.8	29.7	31.8	28.5	28	30.5
30.8	28.1	30.1	29.5	30.4	28.7	29.5	28.5	32.4	27.6	28.5	31	28.5	28.1	30.6
29.9	28.1	30.2	29.1	29.4	27.1	28.7	28.5	32	26.6	28	29.9	28	27.9	30.4
28.8	25.2	28.1	28.3	29.2	27.8	27.3	29.4	30.5	26.7	29.3	31.2	28.1	27	29.4
31	28	30.4	29.4	30	28.7	28.9	29.3	32.2	27.6	29.6	32.1	28.4	28.2	30.4
29.8	26.5	29.6	28.8	29.8	27.4	27.8	28.4	31.3	27.1	29.3	31.9	27.9	27.5	29.7
30.6	27.3	29.4	28.7	29.5	28	27.7	28.6	31	26.9	28.9	31.6	27.5	27.4	29.9
30.9	28	29.5	28.9	29.8	28.6	29.3	28.2	31	27.8	29	31.5	28	27.5	29.1
34.4	34	35.9	34.9	36	34.1	35.7	39.2	36.2	36.1	39.4	34.1	37.7	34	34.5
30.4	30	32.9	33.7	33.4	30.9	32.8	32.5	32.3	33.4	34.5	34	36.9	31.5	34.4
33.7	32.7	34.2	34.9	34.8	32.8	36.2	36.1	34.6	36.2	38.9	34.5	36.7	31.7	35.2
31.4	32.4	33.1	33.9	32.9	31.2	33.8	33.8	33.3	34	36.2	33.9	36	30.3	35
34.6	32.2	34.3	33.9	34	33.8	34.2	39.3	34.6	35.2	36.9	34.1	36.4	32.1	34.1
35.7	33.9	35.9	35.1	36.1	35.3	35.8	40	36.3	36.7	39.6	34.1	37.3	34	34.9
33.7	31.6	35.5	31.8	33.5	33.6	34.2	39.1	33.7	33.5	34.4	32.6	33.9	31.2	33.4
34.2	32.5	33.7	34.5	34	33.9	34.8	38.9	34.7	34.1	39	33.3	36.5	33	33.8
36	33.1	37.1	33.3	35.2	33.1	35.5	38	35.6	34.4	37.2	34.2	36.3	32.6	36
34.9	32.5	35	34.4	34	32.5	33.1	35.3	35	33.4	37.4	33.2	36.9	32	33.6
33.9	32.8	35.8	33.3	34.6	34.2	34.7	38.8	34.6	33.9	37.4	34.3	36.6	32.3	34.7
35.3	33.9	36.3	33.9	36.1	36.8	36.5	39.4	36.6	37.5	38.6	32.9	36.8	32.7	34.9
33.6	32.7	35.8	34.1	34.6	33.8	34.5	39	34.2	34.2	35.9	32.6	35.7	32.1	33.5
35.1	33.6	34.8	34.2	34.8	34.3	33.8	37.6	34.9	34.7	38.1	33.9	37.4	32.3	33.9
35.4	32.8	34.4	34	34.7	33.1	33.6	38.2	32.6	32	35.9	32.3	36.5	30.8	33.1
34.1	32.1	35.5	32.6	34.3	34.3	34.5	39	34.7	33.4	36.3	33.1	35.8	31.7	34.2
34.4	32.5	35.5	32.8	34.9	34.8	34	38.9	35.2	35.1	36.8	33.3	35.9	32.3	33.2
35	32.4	37	32.8	34.2	32.9	34.5	38.1	34.5	34.2	36	33.7	34.6	31.7	34.8
34.1	32.3	34.8	32.7	35.4	33.2	34.4	38.1	34.1	33.2	35.9	32.5	35.9	31.7	33.3
35.4	33.9	36.1	34.7	35.3	34	34.5	38.4	34.3	35.2	38.8	33.7	37.5	32.2	34.2
33.5	33.6	34.6	34.2	33.6	32.7	34.9	35.4	34.7	34.5	37.2	35	36.3	32.2	34.3
35.7	33.4	34.8	33.5	34.4	34.4	34.5	37.4	34.7	34.2	38	33.7	36.8	32.2	34.8
32.4	31	33.5	32.7	33.2	32.3	32.3	37.7	32.4	31.6	34.3	31.5	34.8	30.1	32.1
34.2	32.6	36.1	32.9	33.9	32.2	33.2	37.3	33.9	32.9	36.1	33.4	35.7	32	34.4
35.5	33.1	35.1	34.7	35.2	33.8	34.5	38	34.9	34.7	38.2	33.9	36.1	32	34.3
33.8	32	33.7	34.2	33.7	32.3	35.1	35.3	34.1	36	38	34.6	36.5	30.9	32.9
36.2	34	35.6	35.3	34.4	33.6	34.8	37	34.7	34.6	38.4	34.5	37.3	32.7	34.6
34.7	33.4	36.2	34.6	35.8	35.6	35.5	38.5	35.3	35.2	37.9	33.1	36.1	33.1	34.7
35	32.8	34.2	33.6	34.5	33.8	33.3	36.7	33.7	33	36	34	36.3	31.5	33.8
36.3	34.6	35.4	34.5	36.2	35.1	35.5	37.9	35.2	35.9	39.2	34.6	38.6	33.9	34.6
35.1	33.1	34.5	34	35.1	33.7	34.2	36.7	36.1	34.9	38.5	34	37.8	32.4	34.5
33.6	31	34	31.6	32.6	31.1	32.8	36.1	33.1	32.9	34.8	32.2	34	30.6	33.4

34.6	32.7	35.9	33.1	34.5	34.8	34.8	39.3	35	34.1	35.9	32.7	35.6	32.3	33.4
34.8	33	35.1	35.3	34.8	34.3	34.1	38.4	34.1	34.2	38.6	33.8	36.6	31.8	33.6
34.6	33.6	34.8	34.2	35	34.6	34.5	38.4	35	35	38.9	32.5	36.5	32.5	34
36.4	33.5	34.6	33.5	35.3	34.1	34	36	34.8	34.4	38.5	33.8	37.8	32.3	34.5
35	33.4	35.2	34.8	35.7	34.3	34.7	38.3	34	34.1	37.5	33.4	36.7	32	34.5
34.4	32.2	33.7	31.8	33.5	34.1	33.1	34.6	34.2	32.6	36.2	33.1	35.1	31.1	33.9
36.4	31.8	33.6	31.5	32.9	31.3	33.4	35.3	33	32.1	34.2	32.7	33.2	30.2	33.7
34	31.8	33.7	31.5	32.5	32.8	32.5	34.5	33	32.3	34.9	32.4	33.8	30.2	33.9
33.6	30.7	33	32.5	31.7	31.5	32.1	33.5	33.3	31.8	35.5	33.8	33.9	30.5	33.3
33.8	31.8	32.8	31.1	32.8	33.7	32	34.2	33.3	32	36.3	32.9	34.8	30.5	32.8
35.5	33.2	34.8	34.7	34.4	33	34.8	36.5	35	33.8	38.5	33.5	36.8	31.2	33.3
33.7	32.2	33.6	31.6	33.7	33.3	32.1	33.9	33.6	32.8	36.9	32.5	35.2	30.7	32.3
27.6	25.7	22	29.9	26.2	29.4	26.4	29.8	26.7	26.5	26.6	30.5	30.8	29.1	29.8
31.8	31.1	27.2	33.8	31.6	34.4	30.7	33.5	32.2	30.9	31.5	34.3	35.3	34.3	34.2
29.9	28.4	22.9	30.5	28	31.1	27.5	29.3	27.9	27.7	28	30.9	32.2	31.2	30.5
29.2	27	22.5	31.4	28.4	30.7	26.6	31.4	28.6	29.8	29.5	31.6	33.1	30.1	31.2
27	24.8	20	28.4	24.9	27.9	25.2	27.7	26	26.2	25.3	28.9	29.9	27.5	28.5
28.5	27.5	22.7	30.5	26	29.6	27	29.8	28.3	27.9	27.9	31.3	31.9	29.4	31.1
31.4	29.8	25.6	32.4	29.9	33.1	29.5	31.8	30	29.1	29.9	32.7	33.6	33	32
31.1	30.1	25	33	29.3	33	29.5	33	31.4	30.2	29.9	32.2	32.5	31.7	31.1
26.6	26.7	20.9	27.8	26.4	29	25.7	27.5	25.9	24.9	24	27.2	28.7	28.1	27.2
30.6	27.6	23.5	31.1	28.5	30.7	27.6	31.3	30.4	30.3	29.1	31.7	33.5	30.7	31.5
28.3	27.8	22.7	30.9	27.1	31	26.8	31.2	27.5	26.7	27.4	29.5	31.1	29.6	29.2
30.2	29.7	25.6	32.8	28.9	32	29.3	32.1	29.9	28.7	29.7	31.6	32.3	31.8	32
34.9	30.6	26.3	34.5	31.2	34.8	30.2	34.8	31.9	33.2	31.5	36.6	37.7	34.6	33.7
29.6	27.8	22.9	32.4	29.2	32	27.7	32.4	29.5	30.4	28.8	32.1	33.5	31.3	32.4
20.7	18	14.2	23	17.3	21.8	18.6	20.9	17.8	18.2	19.8	22.7	22.6	21.1	21.8
28.8	28.4	24.1	29.1	27.8	30.5	27.8	28.8	26.2	27.1	27.6	29.9	31	30.4	29.4
28.5	26.2	21.7	30.2	27.3	30.7	26.8	31.2	28.3	29.6	28.3	30.6	32.5	29.8	30.6
26.3	26.2	21	27.5	26.1	28.6	25.1	26.6	24.7	24.5	24.9	27	28.8	27.3	25.5
28	28	23.4	29.6	28	29.9	26.8	28.8	26.6	26.9	26.6	28.9	30.4	29.8	28.5
29.9	27.5	23.7	31.5	26.5	29.9	27.7	29.6	28.2	27.9	28.8	31.2	32.3	30.4	30.9
28	27	22.7	30.9	27.9	30.7	27	31.3	27.5	27.5	27.3	29.9	30.9	29.1	29.6
32	30.5	26	32.9	29.5	32.7	29.8	32.4	30.6	30.2	30.9	33.8	34.7	32.2	33.6
28.5	27.7	23.1	29.8	28.5	32.8	27.4	29.8	30	27.7	27.5	28.8	31	29.4	29.5
30.5	28.7	25.7	32.6	30	32.3	29.5	33.7	29.9	30.6	30.4	32.9	33.2	32.3	31.8
28.4	28.7	23.3	30.7	27	31.4	27.3	29.6	28.5	27.5	27.2	30.8	31.6	30.5	30.3
32	31	27.4	34.7	32.1	33.7	31	34.8	30.9	31.6	31.6	33.6	34.8	33.6	33.2
25.8	25	19.3	26.7	25.1	27.5	24.3	26.2	24.1	24.3	23.1	26.4	28.3	27.7	25.9
28	26.8	22.2	28	27.2	29.1	26.5	28	26.7	26.6	26.5	29.1	30.2	29.4	28.9
29.4	27.7	22.5	31.1	26.9	29.9	27.5	30	28.4	28.6	27.7	31.7	32.5	30.4	31
31.5	30.6	26.1	32.4	30.1	32.8	29.8	31.2	30.2	29.1	30.2	32.6	33.4	32.6	31.9
31.6	30.6	26	33.7	31.2	33.9	30.3	35	31.1	31.6	31.2	33.8	34.3	33.2	33
25.9	25.1	18.9	26.5	24.6	27.6	22.7	24	22.8	23.3	21.5	24.4	25.4	25.6	24.2
31.2	29.4	24.9	33.3	28.9	32.7	28.8	32.8	30.4	29.2	28.5	32	32.9	30.8	30.7
28.8	28.4	22.2	30.4	29.1	30.1	26.4	28.2	27	24.8	24.8	28.7	29.3	29.4	27
30.2	28	24.5	32.2	29.6	31.9	28.9	33.4	29.1	30	30	32.5	33.4	31.7	31.2

18.2	18.1	20.5	17.7	19.2	17	19.5	17.1	16	16.3	17.8	18.8	19	19.8	19.2
20	20.5	18.7	18.9	20.6	20.4	20.5	18.7	18.8	20.5	18.2	20.3	21	21.1	20.9
19.5	21	18.5	18.1	20.4	19.5	19.5	19	17.6	19.2	16.8	19.6	20.4	21.1	20.2
17.8	19.3	18.8	17.8	19.9	18.4	20.2	18.3	17.1	18.6	17.6	19	19.7	20.3	19.7
17.2	18	17.2	16.3	18.3	16.9	18.5	17	15.6	16.8	16.3	17.3	18.3	18.7	18.4
26.1	23.1	26.7	25.8	26	24.2	25.1	26.3	27.7	23.2	24.5	27	25.4	23.9	26.4
25.2	22.6	25.8	25.9	25.9	23	24.7	23.9	26.6	22.2	23.9	26.3	24.6	23.5	26.3
30.6	26.7	31.1	29.2	29.8	26.9	29.6	28.2	31.6	26.8	28	30.7	28.4	27.2	30.3
29.9	27.9	31	28.8	29.9	27.8	29.1	29	31.6	27.4	29.2	31.8	28.4	28.2	31.2
28.5	26.6	29.8	29.9	29.7	26.5	29	28.7	32.7	26.4	26.6	30.6	29.1	27.5	29.5
30.1	27.6	30.7	30.3	30.1	27.6	29.9	29	33.3	27.2	28.3	31.2	29.4	27.8	30.7
31.1	26.9	30.7	29.2	30	27.7	29.9	28.6	32.1	26.7	28.5	30.6	28.5	27.4	30.4
29.9	28.7	30.8	30.7	30.8	27.9	30.2	29.5	33.2	27.4	28.8	31.8	29.8	28.4	30.9
27.3	23.5	26.9	27.8	28.2	24.8	26.6	25	27.6	24.2	25.7	27.4	26.1	25.2	28.4
31.5	28.1	31.6	29.7	29.9	27.2	29.9	29.4	31.7	26.1	27.8	30.9	29.3	27.6	29.8
30.9	26.9	29.1	28.9	29	26.9	29.2	27.7	32	26.1	27.4	30.4	28.4	27.1	29.2
31.3	28.6	30.8	30.2	30.1	27.9	30.5	29.7	33.2	27	29	32.1	29.5	28.8	31.3
29.9	26	30.1	29	29.3	26.6	28.2	26.9	30.3	25.7	26.7	29.1	28	26.5	29.2
31.3	28.8	31.4	30.9	30.8	28.3	30.4	29.4	32.3	27.7	29.5	31.3	30.4	29.4	29.3
28.3	25.3	28.2	26.9	27.8	25.5	26.3	26.9	29.1	25.1	26.4	28.8	26.2	25.4	28.5
29.1	26	28.5	27.5	27.9	26.2	27.2	28.2	30	26.2	27.4	30.1	27.1	26.3	29.5
33.1	30.4	33	30.7	31.7	29.6	31.9	30.9	34.5	28.3	30.2	32.9	30.9	29.9	32.4
29.8	27.2	29.9	28.7	29	27.5	28	28.5	30.5	26.6	28.2	30.3	27.4	26.5	29.8
28	25.5	28.3	27.3	27.4	25.2	26.9	26.9	29.3	25.1	26.3	28.7	26.7	25.1	28.6
27.4	23.9	28	26.8	27.3	24.8	25.3	26.2	28.6	24	26	27.9	25.2	24.6	28
31.8	29.1	31.9	30.7	31.2	28.7	30.3	29.8	32.8	28	29.4	31.9	29.5	28.9	31.7
30.6	28	30.2	28.8	29.9	27.4	29.1	28.2	31.5	26.5	27.6	29.7	28.1	27.4	30
28.3	25.3	28.7	27.6	28.3	26	26.4	26.9	29.8	25.4	26.4	29	26.2	25.3	29
30.1	27.3	31.1	29.6	29.3	27.1	29.9	28.6	32.2	25.8	27	29.9	28.4	26.3	29.6
28	25.5	28	28.5	27.6	25.1	27.6	27.1	29.6	25.4	25.9	28.1	27.3	25.8	27.5
26	24.3	26.9	27.1	26.5	23.6	26	25.5	28.2	23.8	24.5	26.7	25.5	24.5	26
28.3	25.7	28.5	28.3	28.6	25.9	28.7	27.8	29.9	25.8	26.2	29.5	27.6	26	28
32.5	34.1	37.1	31.6	33.4	34.2	33.5	35.9	34.1	33.8	31.7	35.7	32.4	33.5	33.6
32.6	32.6	35	31.6	32.1	32.4	32.4	34	32.3	33.5	31.5	33	31.3	32.4	33.2
32	32.7	36.3	30.7	33.5	32.1	32.3	33.7	31.7	32.9	31	34	30.5	32.4	32.2
30.8	30.3	34.5	28.4	31.7	30.9	31.2	32.1	31.4	31.9	29.6	32.2	29.3	30.1	29.9
32.3	32.6	35.6	30.9	33.2	32.1	32.3	33.2	31.9	31.3	29.9	33.1	30.2	31.6	32
31.1	32	35.2	29.4	32.5	31.3	32.2	32.8	32.1	32.8	30.5	33.2	30.2	30.9	30.9
32.7	34	35.8	31.9	32.9	32.5	32.7	33.4	32.7	31.8	30.7	34.5	31.5	32.5	33.1
33.1	33.5	36.5	32	34	33.1	33.2	34.7	33.3	33.3	31.9	34.4	31.5	33.2	33.6
32.4	31.7	36	30.5	33.7	32.8	33.2	33.6	32.7	33.8	31.4	33.7	30.5	31.8	31.6
31.6	32.4	35.9	30.4	34.2	32.3	32.6	33.7	32.7	34.3	30.7	33	30.8	32.3	31.7
30.5	31.3	32.2	30.2	30.7	29.9	29.9	32.1	31.1	30.6	30.8	31.1	29.5	30.4	31.4
32.6	33.4	36	32.1	33.1	32.8	32.7	33.8	32.8	32.9	31.5	34	31.3	32.5	33.5
32.6	32.6	36.8	31.5	34	32.4	32.4	34.4	32.9	32.7	30.7	35.1	30.7	31.9	32.2
32	33.4	35.2	30.9	32.1	31.8	31.9	32.9	31.6	30.5	30.2	33.5	30.6	31.4	32
31.7	33.1	35.6	30.9	32.5	32.1	31.2	33.3	30.8	31	30.3	33.5	30.4	31.2	32

31.1	33.5	35.1	30.3	32.9	32.4	31.6	33.5	31.8	32.3	29.8	34	29.8	32	31
32.5	33.8	36.8	31.6	34.4	33.2	33.3	34.3	33.4	32.7	31.4	34.5	31	32.5	32.6
32.4	33.1	34.5	31.8	33.2	33	32.5	34.1	32.8	32.6	31.1	33.9	31	33.1	33.4
33.2	33.3	35.2	32	32.9	32.6	32.9	33.9	32.9	31.5	31.5	33.4	32.2	32.8	33.3
32.7	34	36.1	31.8	33.2	32.7	32.4	33.6	32.4	31.9	31.2	34.4	31.4	32.2	33.1
33	33.4	36.3	31.7	34.1	32.9	33.5	34.8	33.2	33.6	31.8	34.9	31.3	33.7	33
32	32.9	33.8	31	31.7	31.3	31.1	32.5	31.9	31.1	30.1	31.7	31.5	31.6	32.5
33.1	34.2	37.8	31.7	34.8	33.3	33.3	34.7	33.2	33.3	31.8	34.3	30.6	32.8	33.1
33.3	33.2	36.7	30.5	35.1	31.9	32.2	33.5	32.9	33.5	30.3	33.3	30.6	31.2	31.2
33.8	34.7	37.9	32.8	35.5	33.6	34.5	35.4	34	33.9	32.4	35.6	31.7	33.6	33.6
34	33.7	36.5	32	33.7	32.8	33.2	34.7	34	34.2	32.2	34.8	32	33.3	33.9
31	31.8	34.1	30.5	32.3	31.5	31.2	33.1	32	32.1	30.6	33.5	29.8	32.1	31.5
33.5	34.5	37.4	33	34	33.5	33.3	34.7	33.2	33.5	32	35.3	32.1	33.3	34
33.4	33.8	36.4	31.2	33.8	31.7	33.1	32.9	33	31.7	30.8	34	31	31.8	31.8
28.5	22.7	23.1	27.7	28.1	27.9	26.9	28.3	28	29.1	29.5	32.3	32.7	29	30.5
30.3	24.5	25.7	27.5	29.8	28.8	28.5	29.5	30.6	30.4	29.9	33	32.2	29.7	31.3
30.7	23.9	25.1	29.4	30.2	30.1	28.5	30.1	31.1	31.6	31.6	35.3	34.6	31.5	33.1
29.3	22.6	24.9	26.2	29.3	25.5	28.4	28.1	29.1	28.8	28.9	32	29.3	27.1	29.9
31.6	24.6	26	30.1	30	30.4	28.7	30.1	30.7	32	32.4	33.8	34.1	30	33.3
32	23.9	26.1	27.8	31.4	28.1	29.6	29.9	30.3	31.8	30.4	34.9	32	29.5	32.4
27.8	22.6	25.4	25.7	28.8	25.6	27.6	27.5	28.5	27.4	28.3	29.7	27.7	26.1	28.8
31.3	24.7	26.5	29.4	31.6	29.1	29.6	30.9	32.1	31.4	31.6	35	32.9	31	33.1
32.2	25.3	25.7	29.3	30.9	30.5	30.4	32.8	31.2	31.3	32.5	34.8	34.4	31	33
31.5	24.5	26.7	29.1	31.1	30.3	28.8	31.3	32.1	31.6	32	33.5	33	30.6	33.8
29.9	23.6	25.7	27	29.3	27.1	28.5	29.2	30	29.7	30.5	32.3	29.4	26.9	30.5
26.2	21.8	22	24.9	25.7	25.2	25.5	28.5	27.8	27.1	28.2	30.1	29.3	25.3	29
29	23.7	25.6	27.2	28.8	27.5	28.3	30.1	29.5	28.6	30.6	31.4	29.5	27.3	30.3
30.6	23.1	24.8	27.8	29.8	28.2	28.4	29.7	30.4	30.9	30.9	34.1	32.1	28.6	33
30.7	24	26.1	27.1	31.1	27.5	29.3	29.8	30.5	29.8	30.2	33.6	30.1	27.6	31.4
32.5	24.4	25.8	29.3	29.8	30.1	30	31.5	31.6	31.8	32.1	33.4	33.8	30.4	33
29.3	23.9	25.2	27.3	29.3	27.4	27.7	28.9	30.9	28.7	30.5	30.9	30.8	28.1	30.3
27.2	21.8	25	26	27.3	25.8	26.3	27.7	28.9	27.4	29.5	30.6	28.2	26.3	29.6
27.4	22.1	24.4	25.4	27.5	26.5	26.8	27.8	28.6	27.7	29.1	30.5	28.9	26.3	28.8
31.7	24.6	26.9	28.9	30.9	29.9	29.9	30.3	31.5	31.2	30.8	34.1	32.4	31	32.4
29.1	23.1	24.6	26.8	28.6	28.1	27.5	30	29.7	30.2	30.9	31.3	30.1	27.9	30.3
30	24.2	26.5	27.2	30.9	27.2	29.4	28.6	30	28.5	29	31.2	29.6	27.4	30.3
29.1	23.1	25.1	26.7	28.4	27.6	27.9	29.5	29.5	28.6	30.6	31.5	29.7	26.9	29.8
29.9	22.9	25.1	27.2	29.5	27.4	27.3	30.1	30.4	30.4	30.7	32.7	32.3	28.5	31.6
28.8	23.1	25.7	26.5	29	26.6	28.4	28.9	30.4	29.5	30.6	32.4	28.9	26.6	30.1
29.2	27.3	31.7	26.4	28.6	27.6	28.9	27.9	29	27.8	27.6	29.3	26.9	27.1	27.9
32.4	31.7	34.6	31	32.7	31.4	31.7	31.8	33.6	33.2	32	33.2	31.3	31.4	32.7
33.5	32	34.8	31.2	32	31.8	33	32.9	34.2	32.8	33.2	34.1	32.2	31.3	32.5
31.7	30.1	33.1	29.1	31.6	29.5	31.9	30.5	32.3	31.6	30.6	32.3	29.8	29.8	30.9
33.2	32.2	35.3	31.3	33	31.1	32.9	31.8	34.8	33.3	32.6	33.2	32	31.1	33.1
31	30	33.5	28.9	31.1	29.8	31	29.7	31.1	30.7	30.3	31.6	29.3	29.4	30.5
31.5	30.5	34.5	30	32.7	29.5	31	29.9	31.4	30.7	30.6	32.3	30.1	30.2	31.3
29.5	29.1	33.1	28.1	31.4	27.6	29.7	29.9	29.9	28.9	28.4	30.4	28.6	28.6	29.4

32.3	30.9	34.2	30.2	32.3	30.7	32.7	30.6	33.2	32.1	31.5	32.4	30.7	30.2	32.3
30.6	29.7	34.1	28.5	31.6	30.2	31.1	31.3	30.6	31.3	29.7	32	29.2	29.8	30.3
31.3	30.4	32.2	30.2	31	30.1	31.2	30	32.7	32.1	30.7	31.6	31.2	29.6	30.4
32.6	31.5	33.8	30.9	31.3	31.7	32.8	32.3	33.8	32	32.2	32.9	31.5	30.5	31.2
31.6	30.3	33.2	29.1	31.6	30.5	31.8	30.8	33	32.5	30.4	32.8	30.9	30.5	31.6
31.4	30.4	32.9	29.5	30.5	29	30.9	29.5	32	31.3	30.6	31.4	28.9	29.3	30.7
33.7	31.7	34.8	31.1	31.8	31.3	33.2	32.2	33.9	31.6	32.2	33.5	31.7	30.8	31.7
34.1	33.9	34.7	36	35.3	36.8	35	37.8	33.7	36.6	35.8	32	32.4	33.6	35.6
28.7	32.7	30.9	34.1	33.8	32.9	31.8	33.8	30.6	33.9	33.7	31.2	31.5	32.6	33.8
34.3	34.4	36.1	36.2	35.5	36.1	34.4	37.5	34.9	37.3	37.4	31.8	35	33.6	35.4
32.4	36.8	35.7	38.3	37.2	35.7	36.3	36.2	34.4	36.6	36.8	35.4	37.6	35.9	38
34.3	34.5	34.9	37.6	36.8	38	35.7	38.4	34.2	38.2	37.2	32.6	33.9	33.9	36.2
34.4	36.8	36.6	38.7	38.1	37.7	37.2	39.6	36.6	38.9	39.3	35.6	36.1	36.2	37.1
30.9	31.8	32.6	33.5	34.3	32.2	32.9	35.9	32.8	33.8	35.5	31.8	34	30.6	31.6
33.7	34.2	36.2	37.3	36.6	36.1	35.1	38.9	34.9	36.5	37.9	34.2	35.7	34.4	34.8
33.3	32.8	33.6	33	33.2	34.3	33.6	35.9	32.7	35.2	33.7	33.2	32.6	33.7	34.1
33.6	36.4	36.3	38.4	37.8	36.8	37.5	38.8	36.5	39.6	39	36	37.1	35.7	37.2
33.4	33.6	34.3	35.8	35.4	35.6	34.7	37.1	32.5	36.5	35.9	32.1	31.8	33.2	34.8
34.7	33.3	36.1	35.7	37.2	35.7	35.3	37.6	34.7	37.1	36.8	31.8	35.4	34.2	36.1
35.3	35.6	36.7	37.6	36.7	37.7	37.1	39	35.6	38.6	37.7	33.6	33.5	34.7	37.4
35.1	32.3	35.4	33.4	33.6	36	33.6	37.3	33.6	37.5	37.5	31.8	34.9	32.1	34.4
36	37	36.7	38.2	37.5	37.8	36.6	39.6	36.4	39.5	38.7	34.2	34.2	35	36.1
37.9	38.5	38.3	39.8	38.2	38.7	38.2	41.1	37.7	40.2	39.9	35.8	36	36.9	39.1
35	36.3	36	38.1	38.3	38.9	38	39.7	35.3	39.2	39.4	35.4	35.5	35.9	36.7
34.2	34.6	34.6	36.2	35.2	35.8	35.1	38.7	33.3	37.2	36.7	33.8	33.8	33.5	35.2
31.6	32.3	33.1	33.6	33.3	31.6	33.3	35	32.4	34	35.6	32.8	34.3	30.8	31.9
31.9	33.9	34.6	35.7	35.7	33.8	34.9	37.5	34.9	36	36.9	33.3	34.4	33	34.5
32.4	32.7	33.5	33.3	33.1	32.4	34	36.9	33.9	33.8	36.5	32	33.9	31.2	32.9
30.7	34.9	33.3	36.1	35.5	34.5	35.1	36	33.3	35.7	36.7	34	34.4	34.3	35.2
34.6	34.2	34.4	36.6	35.7	37.1	35.3	36.8	34.4	37.3	36.4	32	34.8	33.3	35.7
32.9	32.4	33.7	34.2	33.7	33.8	34.6	36.4	33.2	35.6	34	32.4	32.5	33.4	34.7
34.2	33.7	33.3	34.5	35.4	35.8	35.3	36.2	33.5	36.2	34.7	33.9	33.2	34	35.5
32.1	32.2	32.6	33.4	33.1	31.8	34	35.3	34	34.8	36.9	34.2	36.4	31.9	34.9
32.3	32.7	34.7	35.1	35.4	33.2	32.7	37.4	33.7	35.3	37.6	32.7	36	32.2	33.4
33.3	33.1	33.9	32.4	33.5	33.7	33.7	37.5	33.1	34.1	33.9	33.5	33.4	32.7	34.3
32.4	31.9	32.3	34.8	34.1	36	33.5	35.9	32.5	36.1	35.5	30.4	32.1	31.9	34.1
34.4	34.8	36.5	37.4	37.6	37.1	36.4	38.8	35.1	37.5	38	34.9	35.1	35.3	36.5
34.4	33.3	36.2	34.8	35.7	35.6	33.9	37.8	35.5	36.8	38.9	32.4	36.1	34.6	34.4
35.5	36.1	35.8	37.2	37.3	37.8	37.9	38.6	35	38.1	37.8	35	34.6	34.9	36.8
35.3	34	36.2	35.7	36.9	35.9	35	38.4	35.8	36.8	39.5	33.1	36.7	34.3	35.1
31	32.6	34.3	33.6	34	32.5	33.7	35.3	32.8	34.2	36.1	33.3	35.3	31.3	33.4
33.7	33.6	34.6	36.5	34.4	37.1	35.3	37.4	33.1	36.9	35.8	32.1	32.3	33.2	34.9
33.3	33.4	33.7	35.6	34.7	35.2	34.3	37.9	33	36.4	35.3	32.1	33.2	32.5	34.5
35.5	34.8	36.1	36.6	36.6	37	36.6	37.8	34.5	38.2	36.6	31.8	33.2	34.1	35.1
32.8	32.3	32.9	34.9	34.1	35.8	33.7	35.7	32.7	35.6	35.2	30.5	31.4	31.9	34.8
34.3	33.4	34.3	34.5	34.4	36.2	34.8	38.6	34.5	35.6	35.4	32.6	34.8	33.3	34.2
33.2	32.9	32.8	35.1	34	35.3	34.7	37.2	33.3	36.1	34.1	32.3	33	32.7	34.4

34	33.4	34.6	34	34.4	35.8	35	37.8	34.1	35.9	35.3	33.1	34.8	33.3	34.4
33.6	32.6	33.8	35.4	35.2	36.7	36.3	38	33.9	37.2	34.5	33.6	32.6	33	35.2
36.6	34.5	36.6	35.3	36.7	37.7	36.4	39.7	36.2	36.2	37.3	34.9	36.1	34.2	36.1
34.9	32.4	37.2	33.6	34.7	33.9	34.7	40.1	35.5	34.5	36	33.7	35.1	32.9	34.4
34.2	32.1	34.9	32.1	34	35.8	33.5	39.1	34.7	35.1	35.4	32.4	35.8	31.3	33.5
34.9	34.7	36.3	35.2	35.9	35.2	35.9	39	35	36.4	39.2	32.7	37.2	33.8	34.8
35	38.6	36.7	40	39.2	39.8	39.2	40.5	37.9	40.2	38	35.7	36.4	37.2	37.4
35.2	33.2	36	34.9	35.5	36	34	38.1	34.5	35.6	37	32.3	35.2	33	34.5
33.7	31.9	34.7	31.9	33.6	32.6	34.2	37.9	34	34.1	34	32.7	33.5	31.2	33.7
24	22.6	23.3	25.5	25.1	27.2	24.8	26.7	25.2	25.3	24.8	28.6	30.1	26.2	27.7
29.6	27.5	26.6	32.1	28.1	31.1	29	31.1	30	32.1	31.9	34	33.8	31.1	32.7
32.7	30.3	27.4	33.4	30.4	32.9	30.1	32.6	32.4	34.8	32.6	35.2	36.3	32.6	34.2
31.8	30.7	31.8	34.4	32.5	34.4	31.8	33.4	31.1	34.4	31.9	35.2	36.7	32.8	34.7
33.3	30.9	27.5	33	31.4	32.4	29.6	33.8	32.8	34.5	33.4	35.1	36	33.3	34.9
33.4	31.7	31.8	34.3	32.1	33.5	31.5	32.6	30.6	33.4	32.2	35.4	35.6	32.9	35.2
32.4	29.6	30.5	34.4	31.5	34.2	32.1	33.4	33.2	34.1	34.2	36.1	37.2	33	35.4
30.6	28.2	25.2	31.9	29	30.8	28	32.2	30.7	32.8	31.6	33.5	34.7	30.5	31.9
32.7	30.1	27.7	32.5	29.6	32.1	30.2	32.5	32.6	34.2	33.8	35.7	36.5	32.9	35
38.3	36.3	36.3	38.9	37.8	39.4	37	38.3	35.4	38.1	37.5	39.9	40.3	37.9	39.9
28.2	26.4	25	29.7	27	29.4	27.6	28.1	26.3	30.4	28	30.9	31.4	28.7	30.1
32.9	31	27.8	33.8	31	33.8	31.4	33.3	32.3	34.5	33.7	34.8	35.8	32.2	34
34.1	32.1	33.7	35.7	33.9	36.2	33.8	35.2	33.6	35.9	34.8	37.4	37.9	35.2	36.9
33.3	30.5	28.1	34	30.9	32.5	31.7	34.2	32.7	34.6	34	35.7	36.7	33	34.8
29.2	26.7	24.5	29.6	27.1	28.7	26.9	29.4	28.4	31	30.2	31.3	32.7	28.6	30.4
34.9	32.1	32.1	36.3	34.2	34.7	32.1	34.1	33.5	34.3	33.8	35.5	36.4	33.7	36.1
30.6	28.5	27.4	32	29.4	31.2	29.7	30.8	29.8	32.3	31.3	32.9	33.9	30.1	32.3
32.4	30.4	30.3	33.7	31.2	33.1	30.8	34.2	30.6	32.9	31.8	34.5	35.1	31.6	34.5
37	34.6	34.8	37.5	36.3	37.8	35.5	36	34.1	37.2	36	38.6	39	37	38.2
30.1	27.7	28.2	31.4	29.1	31.2	29.4	31.3	30.5	32.2	32	34.3	34.2	29.8	31.5
30.6	28.2	30.1	32.3	30.6	32.4	30.2	31.5	30.1	32.3	30.7	32.9	34.4	31.2	33
32.7	29.9	26.2	34.4	31	31.6	29.9	33.5	32.2	34.1	33.3	34.8	35.8	32.5	33.9
35.3	33.9	34.1	37.3	34.8	37.3	35.3	36.7	34.4	37	35.3	37.5	39.4	35.9	38.3
32.8	30.8	30.4	34.1	31.6	33.3	31.7	33.4	32.2	34.6	33.1	35.5	35.9	32.7	34.3
31.3	29.6	30	33.2	30.8	32.9	30.6	31.6	29.5	32.2	31	34.1	34.9	31.6	33.8
30.2	28.6	27.4	31.5	28.6	30.8	28.3	30.4	28.7	30.9	29.5	32.9	32.7	30.4	32.3
36	33.9	35.3	37.9	35.7	38.2	35.6	37.2	35.4	37.7	37.2	38.9	40.3	36.8	39.4
35.9	34.4	35.7	37.9	35.6	38	35.6	37.3	35.4	37.5	36.1	38.4	40.4	36.7	39.5
22.5	21.8	19.9	24.5	20.7	24.1	21.3	23.1	21.9	24.6	23	26.3	26.8	22.8	25.2
31.9	29.9	28	33.5	30.4	32.8	32.5	32.2	31.1	32.3	31.7	34	35	31.7	33.3
32.5	29.9	29.9	33.3	29.9	32.3	29.7	31.3	30.8	33.3	31.8	34.4	34.6	31.6	33.8
35.2	32.9	33	36.4	32.1	34.5	32.6	34.3	32.8	34.8	34.2	35.8	37.6	33.6	36
28.5	27	28	30.4	28.7	30.2	28.2	29.6	28	30.5	29.2	31.3	31.6	27.8	30.1
32.9	31.5	31	33.8	31.8	33.9	31.7	33.1	31.2	33.6	32.4	35.7	36.5	33	34.8
26.2	23.9	21.8	26.6	24.2	26.6	23.4	26.3	26.7	28.1	27	28.8	29.5	25.9	27.4
33	29.6	30.1	33.8	31	33.3	30.2	32.4	31.8	33.7	32.4	35.4	34.7	32	34.2
31.1	29.9	29.8	32.7	30.5	32.8	30.4	31.6	30.8	32.7	31.2	34	34.5	31.4	33.7
30.2	28.9	29.1	31.5	29.2	31.6	29.3	30.8	30.4	31.6	31	34.2	33.9	30.5	32.2

31.8	30.1	29.5	33.1	30.6	32.4	29.9	31.9	30	33	31.1	34.1	34.3	31.4	33.3
30.8	29.9	30.3	32.5	30.6	32	29.9	31.8	30.6	33.1	30.9	33.5	34.4	30.6	33.1
29.9	28.4	31	29.7	30.4	27.2	30.1	29.3	32.3	27.7	27.7	30.4	28.9	28.7	31.3
32.5	33.3	34.5	32.7	33.2	31.8	32.6	33.5	34.5	30.8	30.9	34	31.7	32.1	33.6
30.3	30.8	32.2	30.4	30.8	29.1	31.2	30.8	33	28.4	29.2	31.6	30.2	29.4	30.8
29.2	28.5	30.7	27.7	28.7	26.5	28.1	28.2	28.9	25.4	26.3	28.3	27.2	27.3	27.9
31.1	31.6	32.5	32.1	32.6	29.4	30.6	30	31.5	28.4	28.6	31.5	30	30.1	31.2
29.2	29.8	32.1	29	30.3	27.8	30.9	31.2	33.8	27.4	28.6	30.8	29	29	30.3
26	25.4	27.9	25.1	25.8	23.8	26	25.6	26.4	24.5	23.9	25.9	25	24.9	26.4
34	32.1	34	32.4	32.9	30.4	32.6	32.5	34	29.2	30.4	33.8	31.8	31.1	32.8
31.3	30.9	33.1	31.1	31.7	29.4	31.2	31.3	32	28.7	29.2	32.3	30.2	30	31.8
30.7	28.8	31.5	29.7	31.3	28.8	30.3	31	33.9	27.3	28.7	30.6	29.3	28.8	30
29.9	30	31.9	30.2	30.1	30.4	32.3	30.2	32.4	28.7	28.6	30.8	29.3	29.5	30.5
30.6	30.7	32.4	31.2	31.4	29.2	31	31.4	30.9	28.7	28.7	32.3	29.3	29.6	31.8
31.6	32	34	31.4	31.8	30.8	31.3	31.8	32.5	29	29.1	32.2	30	29.6	31.7
30.7	30.3	32.3	30.3	31.8	28.8	30.4	30.7	32.8	27.9	29.1	31.1	29.7	29	31.5
30.4	30.5	31.5	29.1	29.9	28.4	29.9	30.4	30.7	26.9	28	30.4	28.1	28.9	30
32.8	32.9	34.5	32.5	32.6	32.5	32.5	33.3	33.2	30.2	30.7	33.3	31.7	32.2	33.1
29.9	30	32.8	30	30.6	28.6	30.5	29.7	32.2	27.7	28.1	30.6	28	28.5	30.2
30.1	28.3	31.1	28.5	29.3	26.9	28.8	28.1	29.3	25.9	26.8	28.6	27	26.7	27
31	30	33.1	30.6	32.1	28.8	31	30.5	31.3	29.8	29.3	30.9	29.2	29.8	30.4
27.4	23.7	27	27.9	28.6	25.6	26.8	26.3	29.4	25.4	25.6	27.6	27.1	26.2	27.8
23.9	21	24.2	24.8	25.2	22.6	23.3	22.7	25.4	22.3	22.3	23.1	23.1	22.7	24.4
27.4	25	28.1	27	28.9	26.1	27.3	27.5	29.8	25.9	26.9	26.9	28.2	25.9	29.5
26.8	23.3	26.9	27	28.5	25.2	26.1	26.2	29	25	25.6	27.3	26.4	25.5	27.4
26.2	23.9	26.9	27.4	27.6	25	26	26.2	28.1	24.5	25.3	26.6	26.5	25.2	27.5
26.5	23.3	26.2	27.1	27.9	24.9	24.7	24.8	27.1	24.2	24.9	26.9	26.2	25.3	27.6
28.2	24.8	28.4	29.3	29.1	26.1	27.2	26.7	30.4	25.9	26.3	28	28.1	26.6	28.8
27.2	24.2	27	28	27.8	25.3	25.9	26	27.7	24.5	25	26.4	26.2	25.4	27
26.8	25.9	21.7	32.5	27.5	29.5	27.2	31.1	26.8	28.1	28.6	30.9	32.5	30.4	29
27.2	26.1	21.1	30.2	26.4	29.6	26.5	30.7	26.2	27.4	27.1	29.5	30.6	29.1	28.5
30.9	29.7	26	32.9	30	32.3	29.5	32.9	29.4	30.4	30.3	32.3	33	31.4	31.1
24.2	22	16.6	25	23.3	26	22.6	26.4	22.5	20.8	22	25.5	26.8	26.2	23.9
25.2	25.7	21.1	26.8	25.5	27	24.7	25.7	24	24.4	23.5	25.9	27.3	27.3	25.9
29.1	27.3	22.9	32.2	28.9	30.7	27	31.8	27.1	28.6	29.4	30.7	32.2	30.2	29.2
30.8	29.6	25.6	33.7	30	32.8	29.6	34	30.2	31.1	29.9	32.9	33.4	31.9	31.9
32.3	30.4	27.2	34.1	31.6	33.6	31.6	36.1	30.8	31.7	31.8	34	34.8	33.9	33.5
30.2	30	24.7	32.3	29.4	31.9	28.5	33.1	28.9	28.9	29.3	31.7	32.2	31.5	30.4
28.5	27.1	21.9	31.4	27.4	29.9	27.5	31.8	26.8	27.9	28.1	30	31.7	29.7	28.7
28.1	27.2	22.6	31.3	27.2	30	27.3	32	27.3	28.5	29	29.4	31.4	29.5	29.1
28.5	26.3	23.1	31.2	28.3	29.2	26.5	31.3	26.6	28.3	28.7	29	31.5	29.6	28.9
26.8	26.8	22.4	27.1	26.5	28	25.8	26.9	24.6	26.1	24.3	27.1	29	28.4	26.7
30	28.7	23.3	31.9	27.8	31.4	28.5	32.2	28.5	28.6	28.9	30.3	32.2	31	30.9
21.8	21.2	19.2	22.5	22.4	22.7	20.8	21.4	20.1	20.7	20.1	22	23.8	24.2	24.2
27.8	25.7	21.5	30.3	27.2	29	25.9	29.9	26	27.2	27.5	30.5	30.9	29.8	28.5
24.2	23	19.2	25	24.2	25.8	22.8	24.7	21.9	22.7	22.6	23.7	26.3	25.7	25.1
20.9	21	19.1	21.7	21	21.3	21	21.6	19.8	20.7	20.7	21.1	23.2	23	22.2

27.2	25.9	21.2	30.8	26.8	30	25.2	31.5	27.2	27.1	27.4	29.5	30.4	29.4	28.3
28.2	26.5	22.7	30.8	28.2	29.6	26.4	30.6	27.3	27.6	28.7	29.5	30.9	30.2	28.8
29	26.9	21.8	32	28.5	30.3	27.6	31.7	27.5	28.6	28.9	31	31.6	30.7	29.5
29.1	27.6	23.6	31.5	28.9	31.2	28.2	33.3	29.4	29.5	29	31.9	33	31.4	31
26.8	26.8	20.1	27.2	25.5	28.8	24.1	28.4	25.2	25.8	25.6	28.5	29.4	28.4	26.8
27.8	27.3	23.4	30.5	28.9	31.4	27.4	31.7	27.4	28.2	28.6	29.8	31.2	30.4	29.1
24.8	24.5	21	26.2	25	25.8	24.4	25.4	23.1	23.6	23	24.1	26.9	26.5	25
23	22.2	20.2	24.3	23.5	23.8	22	24.4	21.5	22.5	21.6	23	26	24.9	23.3
23.8	22.7	18.9	25.3	23.7	25.3	23.1	24.2	21.5	21.8	21.8	22.9	25.7	25	23.2
21.5	20.2	15.5	22.7	21.5	24.1	19.9	22	21.1	20.2	20.1	21.2	23.6	23.8	21.2
25.2	23.4	19.6	25.7	24.4	26.2	23	25	22.5	22.9	22.8	24.1	26.9	26.1	24.3
25.6	25.2	21.1	26.7	25	26.6	24.7	25.7	24.4	23.7	22.5	24.4	26.5	26.7	25.1
24.3	23.3	20.7	24.8	24.3	25.1	22.8	24.2	23.1	22.8	22.1	23.9	25	26	24.1
20.1	19.8	18.5	20.8	20.7	20.8	19.7	20.1	18.8	19.1	18.6	19.4	21.6	22	20.9
26.8	27.4	21.4	30.6	26.3	29.3	26.3	30.7	26.2	27	27.9	30.2	29.9	29.5	28.4
30.1	28.1	24.1	33	29.1	31.4	28.1	33.4	28.2	29.3	29.4	31.6	32.7	31.2	30.6
31.8	30	26.5	33.4	30.5	31.9	29.1	34	29.5	30.4	30.8	31.8	33.1	33.1	32.1
29.9	27.6	22.9	31.3	28.1	30.8	27.6	32.2	28.3	28.8	28	30.8	32.4	30.4	30.3
22.4	22.5	19.3	22.4	23	24.5	21.7	22.1	20.5	21.2	20.1	22.7	23	23.2	23
18.3	19.7	17	19	19.9	21	18.4	18.5	17.6	18.3	17.6	18.2	19.6	20	18.9
22.3	22.3	20.3	23	22.6	22.4	22	22.6	20.5	21.2	20.6	21.7	23.7	24.1	21.8
25.1	23.4	19.6	26.6	25	25.5	23.4	25.2	22.2	22.8	22.6	23.8	26.5	25.1	23.8
24.4	23.7	20.8	25.4	25.6	25.5	23.1	26.7	21.8	24.8	22.4	23.7	26.2	25.3	24.2
18.4	19.8	18.3	18.3	19.8	19.3	19.3	17.9	17	17.7	17	18.4	19.1	19.6	18.4
21.4	22.5	18.9	21.3	22.2	23.5	21.9	21.3	20	21.2	19.4	22.2	23.8	23.6	22.2
20.2	21.3	18.7	19.3	21.7	21.6	21.4	19.7	18.6	19.9	18.3	20.9	21.3	21.9	21
25.2	22.8	24.9	24.7	27.2	24.4	25.4	27.5	28.6	25.4	27.7	28.9	26.6	25.5	28.5
24.8	22.6	24.2	24.7	26.4	24.2	24.5	27.4	28.1	25.9	27.4	28.4	25.6	25.2	27.6
26.6	23.8	26	25.3	28.4	25.2	25.2	27.6	28.4	25.5	28.5	28.8	26.5	25	28.6
23.4	20.2	23.1	22.8	24.7	21.9	23	25.3	26	23.5	25.6	27.7	25.1	24.3	27
25.2	22.8	25.3	25.2	27	24.5	25.5	26.5	28.2	24.8	27.4	29.4	26.4	25.1	28.4
29.3	24.9	27.2	26.9	29.3	26.5	28.1	28.2	30.3	27.5	29.9	31	28.1	26.4	28.8
29.6	25	27.5	28.1	30.3	26.9	27.6	29.2	31.3	26.2	30.4	31	28	26.6	28.3
27.1	22.9	25.3	25.2	26.5	24.6	25.1	26.6	29.8	24.5	28.3	28.2	26.2	25	26.3
24.9	22.5	24.6	24.4	26.1	24.1	24.8	27	27.4	25.2	26.9	28.2	26.1	25.3	27.8
28.2	24.3	26.3	26.3	28.6	26.2	27.2	28.7	30.4	26.5	29.1	30.3	27.9	26.3	29.2
26.8	24.8	26.5	26.3	28.6	25.9	26.9	28.5	29.8	26.6	28.6	30.5	27.5	26.5	29.2
27.2	24.3	26	25.8	28	25.7	26.3	27.7	29.1	26.2	28.8	29.6	27.3	25.8	28.8
25.6	22.8	24.8	25.4	26.8	24.5	25.1	27.1	28.7	26.5	28.2	28.1	26.4	25.8	28.7
25.5	23.5	25.4	24.9	27.7	24.9	25	26.9	27.6	24.9	28.2	28	26.2	25.2	27.6
26.9	24.1	26	25.9	28.5	24.8	25.9	27.4	28.7	25.2	28.3	29	26	24.8	27.8
26.1	23.8	25.1	25.2	27.3	24.4	25.6	27.6	28.2	26	27.4	28.3	26.4	25.7	27.9
25.9	22.9	26.4	25.4	28	25.2	26	27.8	28.7	25.8	28.4	28.9	27	25.7	28.4
26.8	23.5	24.9	24.8	28	25	26.1	27.7	28.8	25.4	27.7	28.2	26.2	25.6	27.9
28	24.1	25.8	25.4	28.1	25.9	26.3	27.3	29.4	26.4	28.6	29.1	27.2	25.6	28.2
27.8	24.1	26.5	26.4	29	25.2	26.5	27.5	29.4	26.2	28.6	29.6	27.1	26	28
24.8	21.9	24	24	25.8	23.3	24	25.9	27.5	23.3	26.2	27.5	25.3	22.8	25.9

26.4	22.4	24.5	24.1	26.8	24.8	24.8	25.9	27.9	25	27.4	27.4	26	24.6	27.4
29	25	26.3	26.1	29.3	26.7	27.1	28.1	31.4	27.6	30.1	30.6	28.6	27	29.5
31.1	28	30.7	30.2	29.1	27.2	30.4	29.7	33.1	26.5	27.6	29.7	28.4	28.9	30.1
28.8	27.9	30.3	28.3	28.5	26.5	28.8	28	29.7	26.1	26.8	28.5	27.4	27.3	28.6
32	31.5	33.8	32.5	32.7	29.9	31.5	30.8	34	29.6	30.1	31.8	30.4	30.2	32.1
29.5	27.9	30.2	29	29.9	27.3	29.1	28.7	31.3	27.1	27.9	30.1	28	27.6	29.5
29.2	27.6	30.1	28.1	29.3	26.3	28.6	27.8	31	26.3	27.4	28.9	28	26.7	29.2
27.2	24.5	26.9	24.9	25.8	23.5	25	25.3	26.5	23.1	24	25.6	24.2	23.8	25.7
30.7	28.6	31.7	29.9	30.8	28	29.4	29.1	32	27.4	27.9	30.9	28.8	28.5	30.1
29.7	28.1	30.6	29.2	29.7	27.1	28.9	28.5	31.4	27	27.8	29.6	28.6	28	30
30.8	28.8	31	29.2	31	29.1	29.6	29.3	32.5	26.9	28	30.7	28	27.4	30.5
30.1	27.8	29.8	28.7	29.5	27.3	28.7	28.1	31.7	26.5	27.8	30.8	28.1	27.5	30.3
29.5	29	31	28.9	30.1	27.1	28.9	28.6	30.3	27.3	27.4	28.4	27.8	27.6	28.5
30.3	28.3	30.7	29	29.7	27.2	28.6	28.1	32.1	26.8	27.2	28.8	27.7	27.1	29.5
31.7	29.8	32	30.8	31.7	28.6	31.1	29.9	34.2	31.2	30.6	30.8	29.7	30.6	31.5
22.7	22.2	21.1	25.5	26	26.9	25.1	28	27.9	28.4	27.5	28.8	29.5	25.9	29.5
27.8	25.1	23.9	28.8	26	28.7	26.6	28.3	27.7	29.7	30.2	31.5	32.2	28.4	30.5
21.8	17	14.7	20.8	18.9	22.2	19.1	23.3	21.3	23.1	21.8	23.9	25.9	20	22.5
32.6	28.1	26.7	31.2	31.3	32.2	30.1	33.8	33.3	34.5	34.2	34.6	35.6	30.8	34
27.4	21.4	19.4	26.2	23.9	27.7	24.6	28.4	26.8	28.7	27.2	29.4	30.1	25.9	28.4
28.8	27.1	25	30.6	28.1	30.5	28.5	30.6	30.4	32.3	31.9	33.4	34.3	29.2	32.4
23.2	20.2	16.6	25	21.5	24.3	21.6	25.2	23.6	25.2	24.5	25.9	26.7	22.8	25.1
26.1	23.2	20.2	26.6	23.5	25.7	23.5	27.1	25.8	28	27.2	28.1	29.3	25.2	27.5
26.9	24.6	22.6	27.3	24.5	26.7	24.7	26.7	25.9	28	26.9	29.1	29.5	26	28.6
26	22.5	18.4	26.4	23.5	26.5	23.2	27.4	25.6	27.3	26	27.6	29.2	25.2	27.3
27.5	25.4	26.8	28.1	27.8	27.9	28.5	29.2	29.3	31.1	30.7	32	33.1	27.3	31.1
30.4	24.5	23.6	30.3	29.2	30.9	27.7	32.7	31.3	33.1	32.9	34.2	34.3	29.9	32.4
29.2	24.5	22.4	28.3	28.1	30	26.5	31.1	29.7	30.6	30.5	31.8	33.5	28.2	30.9
25.5	21.9	20.6	26.5	24	26.8	24	27.4	26.5	27.7	27.4	28.5	29.9	24	27.3
26.3	25	26.3	28.2	26.1	28.4	27.7	28.2	28.4	29.9	30.4	31.3	31.3	26.9	30.5
33	27.7	27.2	32.3	31.6	33	31	35	33.3	35.2	34.5	35.4	36.3	31.9	35
26	24	25.2	27.4	25.4	28.2	27	27.7	27.8	29.4	29.4	29.7	31	27.2	29.6
29.2	25.4	24.9	29.4	27.9	29.9	28.6	30.8	30	31.3	31	31.8	33.2	28.2	31.9
31.4	28.2	28.5	30.4	31.7	31.9	31.4	32.7	32.1	33	33.6	34.4	34.7	31.3	33.6
28.3	26.2	26.8	28.3	28.4	29.6	29.2	30.2	30.3	32.5	31.9	33.2	33.4	28.5	32.1
31.3	25	22.5	28.6	29.2	28.9	27.5	29.5	28.6	31.5	30.7	34.1	32.9	29.4	32.7
28.1	24.5	25.4	27.4	28.2	28.8	28.5	29.5	28.9	30.7	31.2	31.9	32.3	28	30.7
31.2	26.3	26.5	30.6	28.5	29.9	28.5	31.3	31.3	32.4	32.3	32.8	33.3	28.3	32.5
30.3	28.6	26.5	30.1	30.4	30.9	30.1	31.5	31.5	32.5	32.5	33.4	33.7	29.5	32.2
29.7	26.6	28	29.7	29.8	30	30.3	29.3	30.9	32.1	31.5	33.2	33.4	29.1	32.2
27.6	26.3	24	28.6	26.7	28.4	26.7	29	29	30.2	30.5	31.4	33.2	27.2	31.4
29.5	27.1	28.5	29.2	29.3	30.5	30.5	31.4	31.3	33.2	34.2	33.3	33.9	29	32.5
26.9	24.3	23.9	27.4	26.1	28	26.5	28.3	28.4	30	29.5	30.1	30.8	26.4	29.2
29.7	25.4	24.5	30	28.9	29.7	28.1	29.9	29.8	32	32.5	33.9	34	29.4	31.9

2006	2007	2008	2009	2010	2011	2012	2013	2014	2015	2016	2017	2018	2019	2020
34.9	32.3	35.4	30.9	35	34.1	35.5	30.7	31.2	33.9	34.3	34.2	33.6	33.5	34
34.8	32.1	33.5	31.9	34.7	33.9	35.2	30.2	32	33.8	34.7	32.8	32.8	33.5	33.1
34.9	31.1	33.7	31.9	34.2	34.2	33.2	30.9	30.8	33.4	34.1	33.1	33.1	32.9	32.7
34.9	33.9	33.7	33.3	34.6	34.2	34	31.8	32.7	34.5	34	33.2	33.6	33.7	33
33.3	32.8	33.1	32.9	33.7	33.4	32.9	31	32.2	34	33.6	32.6	33	32.9	32.1
34.9	33.1	33.1	31.8	34.5	34	34.8	30.6	32.5	33.8	33.8	32.8	32.9	33.7	33.1
35.5	32.6	34.5	32.5	35	34.7	34.4	30.9	32.3	34.8	34.8	33.9	33.5	34.3	33.8
36.5	33.7	34.2	33	35.2	34.6	35.1	31.4	33.5	34.7	34.7	34.1	33.9	34.7	34.2
33.1	29.9	32.5	30.1	33.4	32.3	33	29.6	29.7	32.4	33.4	31.9	32	32.3	32.3
34.6	32.3	33.3	32.9	33.3	32.9	33	30	31.8	33.5	33.3	32.4	32.7	32.7	32.3
34.7	31.6	34.2	32.1	34.1	34.1	33.9	31.2	31.7	33.5	33.8	33.1	33.4	33.5	33.5
33.5	31.3	32.5	30.1	34	31.7	33	29.4	31.5	33.2	33	32	31.8	32.6	32.1
31.1	28.4	30.2	29.2	31.1	30.7	31.3	27.3	28.2	30.3	31.6	28.8	29.7	30	30.1
34.1	30.8	32.2	29.9	33.1	32.5	33.1	29.5	30.5	30.9	32.2	31.9	31.5	31.8	31.6
34.9	32.6	33.1	31.8	34	33	33.8	30.4	31.9	33.8	34.3	32.9	32.4	32.9	32.8
34.2	30.9	32.2	31.4	33.2	38.2	38.5	32.7	30	32.3	32.5	32.7	33.6	31.8	33.1
34.5	30.4	33.6	30.5	33.6	35.5	37.1	32.4	29.4	34.1	34.8	33.4	34.8	32.5	33.6
33.1	29.1	30.6	30.6	31.5	36.3	35.4	31.4	29	31.2	31.6	31.5	32.6	30.4	31.7
33.2	29.3	32.5	31.5	31.9	36.4	34.6	30.7	28.9	32.6	32.9	31.6	33	29.9	31.4
34.7	31.6	33.4	32.4	33	37.8	37.2	32.4	30.3	33.6	33	33.3	34.3	31.8	33.1
34.6	30.7	33	32.5	32.9	37.9	37.7	32.1	30.1	34.1	33.3	33.7	35	32	33
34.2	30.4	34.4	32.5	33.7	37.4	35.2	32.3	30	33.7	34.4	32.8	34.3	31.6	32.1
34	31.1	34.9	31.6	34.2	35.4	35.2	32.3	30.1	34	33.5	32.9	33.8	31.8	33
33.2	31.2	32	29.6	33.7	34.2	34.5	30.8	29	32.2	32.6	32.5	32.9	31.2	32.9
33.3	30.9	32.6	29.3	33.3	33.8	33.9	31.1	29.1	33.6	33.2	32.2	33	31.4	32.5
34	29	30.6	29.4	32	36.8	36.1	31	28.6	30.8	31.2	31.5	32.8	30.4	31.5
32.5	31.1	32.5	30.2	33	34.6	35.7	30.8	30	32.7	33.2	32.5	33.7	31.9	33.7
33	30.3	31.9	29.6	32.4	34	34	30.4	28.9	31.7	32.7	31.8	33.4	31.3	32.2
32.3	30.2	32.2	29.6	32.1	33.5	35	30.4	29.6	33	33.2	33	33.7	31.8	33.6
33	30.1	32.8	30.5	32.5	33.1	32.7	30.3	29.2	33	32.5	31.7	31.8	30.8	31.9
33.5	34.1	32.7	34.1	32.9	33.2	33	33	34.3	30.6	33.9	33.5	35.5	34.1	33.9
33.7	33.5	31.5	33.5	31.4	32.1	31.4	31.5	32.3	29.7	33.4	32.3	33	33.2	33.3
39	37.8	37.5	39.7	38.1	38.2	37.1	37.3	38.6	37.5	39.3	38.2	38.8	39.3	40.3
33.5	32.9	30.7	34.8	32	33.9	32.1	31.8	32.6	32	34.5	33.1	34	35.2	35.1
40.3	39.7	39.1	41.1	39.8	40	38.8	39.3	39.7	38.7	40.6	39.6	39.9	40.9	41.5
29.2	29.6	28.1	29.3	29.1	28	27.3	27.9	29.1	27	29.7	29.1	29.7	29.2	30.3
33	34	33.1	33.8	33.1	32	31.7	32.6	32.1	30.8	33.4	33.7	33.9	33.8	34.7
37.5	37.2	35.6	39.1	37.8	38.1	36.4	36.8	37.5	36.3	38.8	36.7	37.8	39.2	39.3
27	27.2	26.5	27.9	26.8	26.7	25.9	26.3	26.8	24.2	27	27	27.7	27.7	28.5
40.1	39.4	39.2	41	39.5	39.8	38.5	39.4	39.7	38.8	41	39.7	40.4	40.7	41.3
38.6	39.6	38.7	38.9	38.5	37.2	37.2	38.3	38.4	36.2	38.9	38.8	39.3	38.5	39.1
32.1	32.4	30.9	32.7	32	32.4	31	31.2	31.8	30.3	33.2	31.8	32.4	32.8	33.2
28	28.5	28.4	29.6	28.1	28.2	27.4	27.5	28.5	26.9	28.8	28.6	28.3	29.2	29.7
43	42.1	42.1	43.2	42.1	41.3	40.4	41	41.6	39.8	42.3	42.2	42.4	42.1	43.1
43.2	42.6	42.3	43.6	42.7	41.6	40.6	41.3	42.3	40.1	43	43.2	43.3	43	44.4
33.7	33.6	31.5	34.9	33.4	33.7	32	32.4	33.6	32.1	34.8	32.5	34.1	35.2	34.8

41	40.6	40	41.7	40.7	40.2	39	39.8	40.5	39	41.9	40.8	41	42.1	42.1
34.8	36.3	33.8	35.3	34.3	34.4	33.7	34	35.1	32.3	35.4	34.7	35.6	35.8	35.5
42.4	41.5	41.4	43.2	42.3	41.5	40.5	41.4	41.6	39.8	43.1	41.9	42.5	43	44
39.1	39.3	38.3	40.1	38.9	38.8	37.7	37.9	38.7	37.3	39.8	38.5	39.2	39.7	39.8
36.3	36.3	34.9	36.4	35.8	35	33.9	34.6	35.3	33.3	36.3	35.7	35.5	36	37
37.1	37.4	34.5	38.1	36.1	36.9	36.2	35	36.5	35	37.4	36.2	37.1	38.4	38.6
40.7	40.6	39.7	42.3	41.4	41.1	39.6	40	40.6	39.5	41.8	40.7	41.1	41.8	42.3
33.2	33.2	31.6	33.5	31.2	32.5	31.6	31.4	32.1	29.6	33.4	31.5	32.7	33.4	34.4
37.6	37.2	36.6	39	37.6	37.8	36.3	36.6	37.5	36	38.4	37.1	38.1	38.9	39.3
34.8	32.9	31.9	36	35.7	31.5	32.7	37	37	35	32.4	36	36.7	32.8	35
35.8	34.4	33.8	35.2	34.8	30.8	33	37.1	37.1	33.2	33.5	36.5	36.2	33.1	35
34.8	33	33	33.5	32.4	29.3	31.2	32.5	33.1	30.8	32.5	34.7	35.8	31.3	32.8
34.6	34.9	32.8	32.2	33.7	31.6	33.3	35.6	33.9	30.5	31.1	34.1	34.2	31.4	32.7
35.7	34.3	34	36.2	35.9	31.9	33.4	37	37.6	34.8	33.9	36.6	36.9	33.7	35.2
35.5	35.1	35	35.5	33.4	31.9	33.3	34.3	33.9	30.5	33.9	35.2	36.1	32.5	32.8
33.5	32.8	33.2	33.5	33.1	30.6	32	34.5	34.8	31.7	32.3	33.8	34	30.8	33
26.5	25.2	26.4	26	26.4	24.1	25.5	27	27.3	24.1	26.3	27.3	27.9	25	26.2
37.1	34.5	34	36.1	35.7	32	33.6	36.5	37.5	35	35.2	37.3	37.5	33.7	35.6
19.7	20	18.4	17.5	17	18.5	18.4	18.3	19.5	20.8	18.8	18.6	19.8	18.4	20.1
40	40.6	39.6	40.3	39.9	37.8	38.1	38.7	39.5	36	40.3	40.6	40.5	38.1	39.7
35.5	33.4	32.1	34.5	30.4	31.5	32.7	32.5	32.7	30.1	32.7	33.6	34.7	31.1	30.8
38.8	35.8	36.8	37.1	35.9	34.1	35.5	37.4	37.2	35.8	36.8	38.3	38.4	35.9	36.9
43.8	43.3	42.1	44.3	42.8	41.7	41.6	42	42.6	40.7	43.8	43.7	43.8	43.3	44.2
30.5	30.2	29.2	30.3	28.4	27.2	28.6	30	28.5	25.9	29.2	29.7	29.7	27.8	28.4
35.6	32.4	31.7	33.4	30.1	31.3	31.4	31.4	32.8	31.3	33.8	34.2	35.1	32.6	32.8
39.6	37.7	37.5	38.6	38.4	35.6	36.8	38.7	38.8	37.6	36.9	39.6	39.5	36.7	38.3
38.2	35.2	36.1	36.5	35	34.2	34.5	36.2	36.3	35.3	36.1	37.1	37	34.9	36.3
39	36.6	37	38.6	36.6	35.8	36.2	38.5	38	36	37.6	37.8	38.3	36.6	37.1
42.3	41.5	41	42.5	41	39.8	40.5	41	41.4	39.8	42.1	42.1	42.3	41.7	42.9
24.9	22.1	21.4	21.6	19.5	21.5	20.4	21.1	23.1	22.3	22.8	22.8	23.8	21.9	21.8
39.5	37.7	37.5	38.9	37.4	36.4	36.1	38.6	38.5	36.4	37.8	38.1	38.4	37	37.8
26.1	23.6	22.8	23.9	22	23.9	23.2	24.2	26.1	25.4	26.4	25.7	27.6	24.6	25
36.1	33.8	34.7	34.9	33.8	33	33	35.1	36.2	34.9	34.4	36.3	36	33.9	35.6
36.6	33.4	34	34.2	30.2	31	30.7	33.7	32	30.5	33.5	35.5	35	33.5	31.1
17.8	19.5	16.8	16.4	15.8	17.1	16.4	16.3	17.6	19.4	17.3	17.1	17.1	18.6	17.3
29.5	28.1	27.1	29.6	29.1	25.6	27.5	27.8	29.4	27.6	27.6	29.8	29.4	26.7	28.8
39.5	36.9	37.1	38.3	37.1	36	36.5	38.4	37.8	36	37.8	38.8	39.3	37.2	37.7
36.7	34	34.2	35.2	34.2	32.7	33.8	35.2	35	33.1	35.1	36.7	36	34.5	35.2
29.2	25.8	25	26	22.9	24.6	24.5	25.5	27	25	25.8	28.4	28.1	25.2	24.1
37.9	35.4	35.8	37.5	36	34.7	35.3	36.9	36.4	33.8	36.2	36.4	36.8	35.2	35.8
49.4	49	48.3	49.8	48.6	46.5	46.8	48.3	48.5	45.2	48.2	48.8	49.2	47	47.9
40	37	37.8	39.2	37.7	36.6	37	39.2	38.4	37	38.4	39.4	39.3	37.4	38
38.2	35.2	36.4	36.6	35.4	34.3	35.5	37.2	36.8	35.7	36.7	38.1	38	35.8	36.7
35.7	32.6	33.2	33.2	32	32.2	32.3	33.7	33.6	32.5	33.2	34.5	34.2	32.5	33.1
38.3	35.7	35.9	37	33.7	34.9	35.2	35	35.8	34	37.4	37.2	38.9	36.3	36.3
36.3	34.4	35	35.3	34.3	32.9	33.8	36.1	36.3	35.1	35.1	36.8	36.5	34.1	35.6
44	43	42.9	44.5	43.6	42.5	41.6	42.5	43.3	41.5	44.5	44	44	43.7	45.2

43.7	43.6	43	44.2	43.8	42.2	41.6	42.7	42.5	41.1	43.9	43.9	44.1	43.4	44.9
33.5	32.8	33.1	33.1	32.4	29.6	31.9	33.4	33.3	30.4	33.8	33.3	33.1	30.5	33.1
28.8	27	28	28	27.1	25.1	26.4	28.4	28.4	26.2	27.2	28.7	29.3	26	27.7
35.4	33.1	33.9	34	33.3	31.9	33.3	35.3	35.4	34.2	34.2	35.9	35.5	33.2	34.8
36.3	32.1	33	32.8	32.1	31.9	32.7	34	34	32.3	33.8	35.8	34.7	32.7	33.7
34.2	32.9	31.6	31.4	29.6	28.3	30.6	32	32.2	32	32.5	34.3	33.3	30.3	30.2
27.1	23.9	23.6	24.7	20.8	23.1	22.9	23.6	25.8	24.2	24.9	24.8	26.1	23.9	24.8
29.4	26.9	27.6	26.2	23.6	25.8	26.6	25.9	26.9	27.1	27.2	28.9	29.2	30	27.9
34.4	31.1	30.9	31.6	28.1	30.8	29.1	30.7	31.8	31.1	32.3	32.2	33.7	31.2	31.2
37	34.5	34.8	34.8	33.5	33.6	33	35.5	35.3	34.1	35.1	35.7	35.7	33.6	35
35.4	31.5	32.8	34.2	33	30.8	31.8	34.4	34	32.2	33.1	34.6	35.2	31.7	33.5
31.7	32.3	32	31.2	31.3	29	30.7	32.4	33	30.1	30.9	32.1	32.7	30.2	30.5
30.2	27.6	27.9	27.4	25.5	26.9	27.3	26.9	27.9	28	27.5	28.8	28.5	27.6	26.9
25.2	24	22.8	22.9	21	21.9	22.4	22.2	23.6	24.2	22	23.6	23.1	23.8	23.3
24.9	24.1	23.4	23.1	20.2	22.2	22.2	22.6	25	26.7	23	23.3	24.6	23.3	23.5
29.7	27.5	26.9	25.8	23.6	25.5	26.6	25.9	27.6	26.6	27	29.2	26.8	27.1	26.1
26.5	26.7	26.7	26.4	26.6	27.9	27	26.4	26.8	25.7	28.1	26.8	28.6	27.1	27.2
31.5	33	31.8	31.7	31.9	30.8	31.5	32.4	32.2	28.8	31.4	32	33.9	32.1	32.1
32.5	34	32.8	32.5	32.7	32.2	32.4	32.9	32.9	29.9	32.2	32.9	34.7	33.1	33.1
26.1	27.1	26.3	25.9	26.8	26.9	26.7	26.6	26.7	25	25.9	26.9	28.2	26.7	26.8
36	37.5	36.3	35.7	35.8	34.2	35.7	36.3	36.3	32.9	35.4	36.1	37.7	36	35.4
23.8	25.4	24.5	23.2	24.5	23.9	24.6	23.6	24.5	22	24	24.1	25.9	24.6	24
26.3	27.1	27.6	27.3	27.8	28.4	27.8	27.3	27.2	25.2	27.5	27.3	28.8	28	27.4
31.4	31.9	32	28.4	31	31.6	32.2	29.7	29.4	29.5	31.6	31.6	30.6	31.4	31.5
24.4	26.2	24.9	24.9	24.8	25.1	24.8	25.1	25.3	22.8	24.5	24.1	26.6	25.4	25.1
30.8	31.3	31.8	31	30.5	32.7	32.4	30.7	30.6	29.4	33.2	31.2	31.8	31.6	31.8
33.5	32.8	33.6	29.8	31.9	34	36.1	30.9	31.1	31.9	32	33.6	31.9	32.9	32.5
30.9	31.8	32.7	28.5	30.5	30.8	31.4	29.8	29.6	28.8	31.6	31.5	30.5	31	31.4
31.5	31.7	32.3	29.4	31.4	34.5	35.8	31.2	31.4	32.5	33.8	33.2	31.8	29.4	31.4
25.3	26.5	26.5	26.6	26.2	28.5	27.3	26.8	26.4	24.8	28	26.6	28	27.5	27.6
29.2	30.9	28.6	27.3	27.8	26.3	29.5	29	28.3	25.8	28.8	29.2	30.8	28.3	28.8
29	29.5	30.4	27.1	29.1	30.3	30.1	28.3	28.2	27.1	29.7	28.7	29.5	29.7	29.6
31.8	31.7	33.4	30.3	31.2	33.2	33.8	31	30.3	30.8	33	31	31	33.7	32
33.8	33.2	35.2	33.4	34.4	37.1	37.7	33.7	33.6	33.7	35.3	34.1	34.2	34.9	34.1
32.6	33	34.4	32.6	33	35.7	35.2	31.6	32.7	32.5	33.6	32.7	33.1	33.2	33.8
34.2	33.4	33.8	29.9	32.4	33.3	34.5	30.6	30.9	32.5	33.6	34.5	32.4	34.1	33.6
24.1	25.2	25.5	26.1	25.1	26.1	24.8	25.6	25.8	23	25.1	25.2	26.3	24.6	24.6
31.7	31.9	33.3	30.5	31.7	33	32.4	30.3	30.3	29.2	32.7	31	31.6	32.4	32
34.4	34	35.5	33.2	34.1	36.6	37.4	33.5	33.5	34.1	36.6	35.1	35.4	35.8	35.3
34.3	33.5	35	32.9	34.1	37.4	37.2	33.6	33.1	33.9	34.1	33.5	33.4	34.9	33.8
28.8	27.3	28.6	25.9	29.9	30.8	31.2	31	28.5	28.7	30.7	27.9	30	30.9	30.9
29.5	27.7	29.2	26.6	31	30.7	30	30.1	28.6	29.1	29.9	28.2	28.8	29.4	30.3
27.3	25.6	26.5	24.3	28	27.6	27.4	27.7	26	26.1	27.6	25.8	27.5	28	28.2
27.5	26.6	27.4	24.7	29	28.5	27.8	28.3	26.4	27.2	28.1	26.2	27.1	28.2	28.1
29.4	28.1	29.4	27	30.7	31.3	31.5	30.2	29.2	28.7	30.7	28.9	29.3	30.6	31.5
31.8	30.6	31.7	29.5	33.1	33.7	33	31.5	30.8	30.7	32.3	31.2	31.6	32.7	33.2
30.2	28.7	30.2	28.1	31.3	32.4	31.7	30.2	29.5	29.3	30.5	29.7	29.9	31.4	31.8

31.1	31.2	31	29.6	33.1	33.2	33.3	31.5	30.3	30.6	32.2	31.5	31.4	33	33.2
30.5	30.8	30.4	28.9	32.4	32.3	32.7	30.6	29.8	29.7	31.4	31	30.5	31.9	32.2
32.4	33	32.2	33	33.1	33.2	33.1	31.8	33	33.1	34.4	33.3	33.5	33.3	34
34.4	34.9	33.7	33.3	34.7	34.4	34.1	31.5	33.7	35.5	35.1	34	33.3	34.2	34.1
31.6	32.2	31.2	31.8	31.5	31.7	31.5	30.5	31.4	33	32.7	32.4	33	32.4	32.5
33	33.8	33.1	34	34	34.1	33.9	32.5	33.7	33.9	34.7	34.3	33.9	34.2	34.5
31.4	33.1	31.5	33	31.5	32	30.4	30.6	32.5	32.4	32.2	32.1	33.2	32.1	32.5
32.6	31.5	31	31.2	33	32.3	32.4	31.9	31.8	31.8	33.6	31.9	31.3	31.8	32.6
32.7	32.2	31.3	31	32.5	32.5	32.2	30.1	31.9	33	33.4	31.9	31.3	32	32
31.5	33.5	31.6	33	32.9	32.4	31.8	31.2	32.8	33.5	32.9	32.7	33.1	33.1	33.1
31.7	32.6	31.6	32	33	32.1	32.6	31	31.8	32.9	33	32.7	32.2	32.5	32.3
34.7	35.5	33.4	33.5	34.6	34.1	34.5	31.3	33.6	35.1	34.8	33.6	33.8	34.4	34
32.7	33.1	32.6	33.3	33.2	33.2	33.1	31.6	33	32.6	34.1	33.2	32.8	33	33.5
33.2	32.7	31.5	32	33.2	33.4	33.3	31.2	32.3	33	34.6	33	32.1	32.9	32.8
32.7	32.2	31.3	31.9	33.1	32.8	32.3	32.5	33.7	32	34.2	32.8	32.6	32.6	33.2
31.1	32.5	31.2	32.8	32.3	32.3	31.3	31	32.1	33	32.3	32.5	32.5	32.9	32.5
31.6	31.9	30.7	31.3	32.2	32.1	32.3	30.5	31.4	32.2	33.9	31.9	31.4	32	32.2
32.6	33.2	32.2	32.5	33.1	33	32.9	31.6	32.9	33	34.2	33.4	32.8	33	33.4
32.7	33.1	32	32.8	33.7	33.1	33.6	30.7	32.5	33.4	34.5	32.3	32.5	33	33.4
34.5	33.6	32.9	32.8	33.7	33.3	33.3	31.1	32.9	34.2	34.8	33.3	32.5	33.2	33.5
33.7	32.3	32.4	32.3	32.9	33	32.7	30.1	31.9	33.5	32.8	31.9	32.2	32.3	31.7
32.7	33	32.3	32.6	33	32.7	32.9	31.6	32.9	32.3	34	33.1	32.7	33.3	33.9
33.9	33.3	32.2	32.5	34.2	33.4	33.1	31.5	33.6	33.3	35.8	33.3	32.3	33.9	34
33.3	32.8	32.2	32.6	33.5	32.9	32.7	31.6	32.1	33.1	34.7	32.9	32.2	32.8	33.3
32.2	28.7	31	29.5	32.5	32.6	32.7	28.5	29.2	31.5	33.1	31.2	30.1	31.3	31.4
34.4	32.2	34.2	33.5	34.9	35.5	35.2	31.3	32.9	35.6	35.6	33.3	32.6	34.7	34.5
33.9	32.9	33.3	32.3	34.3	34.5	34.3	31.5	33.3	34.2	36	33.5	33.4	34.5	34.5
34	30.6	32.8	31.2	34	33.5	34.1	29.8	31	33.2	34.5	32.2	32.1	33.5	33.3
34.1	33.3	33.2	32.4	34.2	34.2	33.7	31	33.4	33.7	34.7	33.6	32.7	33.4	34
35.3	33.1	34.4	33.9	35.3	35.4	35	31.7	32.7	34.9	35.7	33.5	33	34.6	33.9
33.7	32.8	32.9	32.8	34.2	34	34.7	32.3	32.8	33.7	35	32.8	32.9	33.5	33.9
33.8	31.6	33.5	33.2	35	34.7	34.7	30.9	33	34.5	35.5	33.3	32.7	34	34.2
34.6	34	33.8	33.9	35.7	35.4	35.6	32.8	34.5	35.4	36.7	34.4	34.1	35.3	34.8
35.6	34.3	35.1	33.1	36.5	35.5	35.2	31.8	34	35.3	36.1	34.4	33.5	35.2	34.7
35.3	34.4	33	31.8	34.3	34.1	34.5	30.2	32.5	33.9	34	32.5	32.6	33.4	33
34.4	33.9	33.1	32.3	34	34	33.9	31.2	33	34.1	34.7	33.2	32.6	33.8	33.5
32.8	30	32.2	31	33.3	33.5	33.5	29.8	30.2	32.5	34.3	32.2	31.3	32.5	33.1
33.8	30.8	32.6	31.1	32.4	33.2	33.5	29.2	30.5	33.1	34.3	32.1	31.4	32.5	33.6
35	32.1	32.7	32.5	34.2	33	34	30.3	31.9	34.1	35	32.5	32.2	33.8	32
34.5	33.6	33.2	32.5	34.7	34.3	34.8	31.1	32.5	34	34.6	32.7	32.1	33.8	33.8
32.6	30	31.8	30.5	32.9	32.3	33.2	28.5	29.6	32	32.3	30.5	30.6	31.7	31.3
33.6	30.8	33.2	32.3	34.3	34.5	34.4	30.5	31.7	33.6	34.8	33	31.9	33.6	33.7
31.7	28.9	30.9	29.5	32.4	32.2	32.5	28.2	29.2	31.3	33.1	31	30.5	31.5	31.8
34	31.5	32.8	32	34	33.1	34.1	29.3	31.8	33.8	34.4	32.5	32.2	33	32.8
32.8	32.4	31.2	30.9	32.4	32.4	32.7	29.5	31.4	32.8	32.8	31.7	31.5	32.4	32.2
33.6	31.2	32.4	31.6	33	33	34	29.7	30.9	33.2	33.4	32	32.2	32.5	33
32.2	30.1	32.7	32	33.1	33.9	33.9	29.8	31.1	32.7	33.6	32	31.1	32.4	33.5

31.3	28.9	28	25.8	28.7	30.4	31	28.4	25.3	27.5	27.7	30.2	28.1	28.2	29.9
30.1	28.8	28.8	25.2	28.4	30.3	31.4	28.7	25.5	27.4	27.7	29.5	27.7	28.1	28.8
31.3	28.8	28.3	26.2	29.2	31.8	34	28.5	27.1	28.4	27.8	31.7	30	30.7	30.3
30.8	28.3	28.6	25	29.1	30.5	32.7	28.2	26	27.8	27.3	30.1	29.3	30.3	30.4
29	27.7	27.5	23.5	27.1	29.1	31.5	27.6	25.1	26.3	26.6	27.4	26.8	28.5	28.7
31.3	30	28.9	27	29.9	32	35.6	29.8	27.2	28.8	28.9	31.8	29.9	30.4	31
30.6	30	28.5	25.8	29	31.3	34.3	28.5	26.5	28.3	28.5	30.4	29.8	29.7	29.6
29.5	28.9	28.8	25.4	28.2	29.8	32.2	29	25.8	27.5	27.8	29	29	29.3	30.1
28.2	27.6	27.2	23.3	27.3	29.7	31.5	27.3	25	26.2	26.4	27.1	27.2	28.3	28.2
30	28.6	28.1	24.9	28.5	31	33.2	28	25.6	27.2	27.7	29.3	28.2	29.4	29.9
31.2	29.4	28.7	26	29.8	32.3	34.4	28.9	27.1	28.4	28.6	30.8	29.4	31.6	30.1
31.7	31.2	29.6	26.6	29.6	31.7	33.2	29.1	26.6	28.8	28.9	30.2	29.3	29.3	30.7
30.5	29.1	27.7	24.7	27.7	29.5	32.9	28.1	25.1	26.8	27	28.2	26.9	27.9	28.7
30.2	28	27.2	25	28.8	31.3	34.2	27.9	25.7	27.5	27.7	30.8	29.1	30.7	29.6
31.7	30.6	29.4	27	28.8	31.1	33.7	29.3	26.6	28.3	29.2	30.8	28.9	28.9	30.8
29.3	28.1	27.7	24.1	27.7	29.2	31.9	27.2	24.8	26.5	26.7	27.4	27.1	28.2	28.6
30	28	28.1	25	29.3	31	33.4	28.4	26.1	27.5	28.6	29	28.5	29.7	29.7
29.9	29.5	27.8	25.3	28.2	30	32.7	28.2	25.1	27.3	27.7	30.1	28	28.3	30.1
29.8	29.7	28.8	26.1	29.6	30.5	32.6	28.7	25.8	27.7	27.7	29.2	28.7	29.3	30.3
32.3	29.1	27	26	29.1	33	34.6	28.3	26.5	27.8	28	31.1	29.3	29.7	29.4
31.8	30.9	30	26.7	29.5	31.3	34.1	29.3	27	28.3	29	30.8	29.1	29.8	30.4
29.8	29.2	28.1	25.4	27.5	29.1	31.7	27.8	25.2	26.8	27.2	28.9	28.2	27.9	29.5
31.1	31	29.9	27.2	29.6	32.2	34.2	29.3	27.8	29.1	27.6	30.4	29.9	30.6	30.5
36.8	37.5	33.2	32.9	31.3	31.7	35.5	36	36.3	33.7	32.8	35.6	34.3	31.7	32.6
35.7	36.8	33.2	33.3	31.8	31.2	33.7	36.3	36	33.8	31	36.1	33.7	31.5	31.7
33.3	34.5	33.1	30.5	31.7	30.3	31.7	32.2	32.9	29.3	31.4	33	33.6	31.6	30.8
30.9	32.6	30	27	28.3	28.3	28.4	30.5	30.3	25.7	29.2	29.9	30.3	28.1	27
30.3	31.2	27.9	28.3	26.6	25.2	28.5	30.1	30.5	30	26.9	31.3	29.3	27.5	29.9
31.8	34.5	30.8	30.1	29	27.8	29	31.4	31.5	30.9	28	33.7	30.5	28	29
32	33.1	31.2	28.9	29.7	28.9	31.5	31.5	30.6	27.2	30	31.3	30.9	29.9	28.8
30.8	32.6	28.7	29.9	27.2	26.9	29.3	29.9	30.1	30.1	26.4	31	30	27.4	27.1
30.5	32.7	29.3	27.3	27.5	27.6	29.3	30.7	29.2	26.7	28.1	30.9	30.2	28	27.5
30	30.4	28.8	27.3	27.2	26.1	28.5	28.3	27.9	24.8	28.2	28.2	28.9	26.9	26.1
32	33.5	29.4	29	27.4	27.4	31.2	32	31.6	29.3	27.5	32.4	30.6	28.2	28.2
29.4	31.7	27.9	28.2	26.6	24.8	28.8	30	29.4	28.9	26.5	31.5	29.2	26.5	27.2
33.3	33.6	31.5	30.3	30	30.3	31.9	32.8	32.1	28.3	31	32.8	32.2	30.5	29.4
35.1	36.4	33.9	33.5	32.4	31.9	35	36.2	35.1	32.2	32.6	35.7	34.9	32.3	32.1
34.4	34.8	32.4	32.3	30.8	31.5	33.1	33.8	33.6	29.6	31.6	35	33.9	31.7	31.5
34.3	35.2	31.4	32.8	30.8	29.2	32.1	34.3	34.6	32.9	29.9	34.2	32.9	30.6	31.2
36.8	36.8	34.5	34.3	33.2	32.4	35.6	37	36.1	33.5	33.2	36.4	35.4	32.7	32.9
39.2	39.9	37.6	36.5	35.7	35.8	37.9	39.3	38.1	33.8	34.5	38.4	38.1	35.6	35.3
34.8	35.1	33.2	32.6	30.7	31.4	34.2	35.1	33.6	29.8	30.8	34.7	33.7	31.4	31
36.8	32.9	33.5	32.8	32.9	31.1	34.5	36	34.7	32.7	31.7	35.2	34.3	31.8	31.7
31.3	32.3	30	29.3	28.1	28.8	30.6	31.2	31	27.7	29.2	31.8	30.4	29.2	28.9
29	29.6	27.8	25.5	24.5	25	26.5	28.3	28.1	24.4	26.3	28.3	28.4	25.3	26.1
31.1	31.5	28.7	29.4	27.9	26.4	28.8	30.6	31.5	30.2	26.8	30.8	29.7	27	27.6
32.2	33.5	31.3	29.3	29.2	30.7	31.9	32.7	32.3	28.7	30.4	32.8	32.1	30	29.8

33.5	33.1	30.9	29.4	29.3	30.2	31.6	33	31.9	29.2	30.2	32.4	31.9	30.8	29.8
32.2	33.2	30.2	29.5	28.7	30.2	31.2	32.2	31.3	28	29.9	32.3	31.6	29.4	29.3
35.4	35.4	32.4	33.1	31.8	29.8	33.7	35.2	35.7	34	30.7	35.2	34.2	31.6	32
31.1	31.8	30.4	28.1	27.9	28.4	29.6	31	30.7	27	29.1	30.6	30.1	28.7	28.4
32.1	32.6	29.1	29.5	28.1	26.8	30.3	32.2	31.3	30.7	27	29.5	31.2	28.3	28.8
30.9	28.7	29.6	27.3	31.3	32.9	36	28.6	27.9	29.4	30.3	31.6	29.9	32.2	31.4
32.1	29.8	30.1	27.4	31.5	33.4	36.5	28.9	28.2	29.6	30	32.4	30.3	31.6	31.5
29.6	28.1	28	24.6	29.2	31.2	32.9	27.7	26.2	27.3	28.9	28.8	28.7	30.2	30.2
31.1	29.5	29.1	27.7	31.6	32.1	35.2	28.6	27	28.3	29.1	31	29.4	30.4	30
29.2	28	28.1	24.6	29.4	30.9	33.4	27.4	26.2	27.1	28.2	28.1	28.4	30.2	29.7
29.5	28	28.2	24.6	29.7	30.8	32.9	27.1	25.8	26.2	27.9	28.6	28.6	30.2	29.8
32.2	29.7	30.6	27.5	31	33	35.5	28.8	28.2	29.5	29.1	31.4	29.3	30.7	30.8
31.7	29.7	29.6	27.2	31	32.9	35	29	27.7	28.8	29.4	31.8	29.8	31.6	30.8
30.9	28.6	29	26.5	30.8	33.1	35.2	27.8	26.6	28.3	29.1	30.4	28.9	30.9	30.6
31.8	30.7	31	28.4	32.3	33.7	36.6	29.2	29	30.8	31.3	31.8	31.4	31.5	32.1
30.1	27.1	27.4	24.8	28.6	30.8	31.9	27.3	25.4	27.2	27.8	28.9	28.2	30	28.6
30.2	28.4	28.6	25.6	29.9	31.4	33.4	27.6	25.8	27.3	28.4	28.2	29.1	30.1	29
30.3	29	29.6	26.8	30.5	31.7	35.3	27.6	27.1	29.1	29	30.1	29.6	30.1	30.5
31.4	29.1	29.8	26.9	30.9	32.6	35.6	28.3	27.4	29	29.6	31.5	30.1	30.8	30.9
30.6	29.1	29.7	26.6	31.4	32.2	35.3	27.4	26.8	28.4	29.1	30.2	29.2	30.8	30.1
29.5	27.8	28.6	25	30.2	31.1	33.3	28.2	26.1	27.9	29.6	27.6	27.9	30.6	30.2
31.8	30.5	30.4	27.8	31.7	33.4	36.1	28.8	28.9	29.8	30.8	32.3	31	31	31.4
30.6	28	28.1	25.1	29.9	31.3	32.5	27.9	25.8	26.9	27.6	28.6	28	30.1	29.4
31.6	29.8	30.2	27.8	31.4	32.3	36.2	28.6	28.1	29.8	30	31.4	30.8	31.1	31.3
30.3	27.7	28.6	25.9	31	31.5	34.6	26.9	26.2	27.7	28.6	29.7	28.7	30	29.6
29.1	29	28.8	26.2	30.6	31.5	33.6	27.2	26.7	28.6	29.1	29.9	29.3	29.1	30.1
31.9	30.4	30.6	28.8	32.1	34.3	36.7	29.1	28.9	31	31.3	32.4	31.5	31.5	32.2
31	29.2	29	26.4	30.2	32.3	34.2	28.5	27.1	28.1	29.2	31	30.2	30.9	30.3
29.8	27.4	28.4	25.3	30.2	31.2	32.6	27.3	26.7	27.5	27.9	28.6	28.8	30.4	30.1
31.5	30.7	30.8	28.3	32.5	33.5	36.2	29.1	28.8	30.7	31	31.8	31.4	31.3	32.1
29.6	28.2	28.9	26.4	30.7	32	33.2	27.3	26	27.5	28.2	28.9	27.9	30.1	29.8
30.8	28.5	29.5	26.4	31.3	32.2	34.9	28.7	27.4	28.2	29.6	30.7	29.9	31.9	31.2
30.8	28.5	28.6	25.7	29.7	31.3	33.5	29.1	26.4	27.8	29	30.1	29.3	30.6	29.7
29.6	28.3	28.6	25.8	30	31.7	34.2	27.2	26.1	27.9	28.3	29.7	28.3	30.2	29.9
30.8	29.2	28.7	25.5	29.6	31.6	33.6	28	26.1	28.2	28.5	30.1	29	30.9	29.4
30.4	28.2	28.8	26.1	30.3	31.3	33.9	27.8	26.4	28.1	28.7	30	27.9	31	30.4
30.3	27.8	28.3	25.6	30.1	31.6	33.9	26.9	26	27.5	28.6	30.2	28	30.5	29.8
29.1	27.1	27.5	24.4	29.3	30.2	33	27.6	25.7	26.2	28	27.8	28.1	29.5	29
30.3	28.5	28.5	25.2	30.1	31.6	32.9	27.5	25.7	26.8	28.1	28.9	28.5	30.9	30.2
31.4	28.5	28.4	25.9	30.3	31.9	34.4	28.2	26.1	27.7	28.4	30.2	29.1	30.5	29.8
31	29.7	29.3	27.9	31.6	32.5	36.1	28.1	28	29.8	30.2	31.4	30.6	30.7	31.3
28.4	27	27.5	25.6	29.2	30.9	31.9	27.1	25.6	26.8	28	27.8	28.8	30.1	29.9
29.9	29.1	29.2	25.8	30.4	31.6	33.4	27.9	27.3	28.6	29.3	29.4	29.6	31	31.1
30.6	29.8	30.3	26.9	31.8	32.6	34.7	28.6	28.5	29.6	30.2	30.6	30.8	31.9	31.9
31.1	30.4	31	27.1	32.6	33.3	35.1	29.4	29.2	30	31.2	31.3	31.3	32.3	32.5
31.6	31	30.8	27.4	32.2	33.8	34.9	29.5	29.1	30.2	31.2	31.2	31.4	32.7	32.8
29.6	27.9	28.5	26	29.7	32.4	32.4	26.9	25.6	27.4	28.4	28.3	28.2	31.2	29.7

27.5	27.6	26.8	24.2	28.8	30	31.6	26.2	24.5	25.6	28	27.4	28.4	28.7	30
29.6	28	28.2	25.4	29.9	31.9	34	27.1	26.2	27.9	28.4	28.5	28.9	30.1	30.5
29.1	27.6	28.5	25.6	30.1	31	32.2	27.2	26.1	27.1	28.8	28.1	29.1	30.4	30.4
31.2	30.1	30.2	27.2	31.7	34.2	35.2	29.3	27.9	30	30.5	30.3	31.1	32.2	32.3
30.3	29.7	29.8	27.1	31.6	32.4	35.1	28.8	28.2	29.5	30.1	30.5	30.5	31.7	31.6
29.4	27.9	28.5	25.8	30.1	32.1	35	27.4	26.5	28.1	28.7	29.1	30	30.7	31
28.9	28.1	28.4	25	29.4	30.3	32.6	27.1	25.9	27.1	28.9	28.2	28.5	29.6	29.8
29.5	28.2	29	26.1	30.6	31.9	33.3	28	26.5	28	28.9	28.8	29.1	30.8	30.7
31.8	30.8	31	27.9	32.8	34.1	37	30.1	29.8	31.1	31.5	31.9	31.7	32.2	32.6
29.2	27.9	27.7	24.7	28.6	30.5	32.7	26.2	25	27	27.4	27.4	27.6	29.5	29
28.2	27.3	27.4	24.6	29.1	31.3	32.1	26.2	24.8	26.6	27.8	27.6	28	29.5	29.3
31.8	30.3	30.8	27.9	31.9	33.3	36.6	29.5	29.2	30.4	30.8	31.9	31.7	32	32.2
32.3	31	30.7	28	32.4	33.2	36.2	29.2	28.4	30.2	30.2	31.1	30.9	31.2	31.8
30.1	28.5	28.9	26	30.8	32.1	34.8	27.4	27	28.3	29	29.8	30.2	30.8	31
29.3	28.3	28.5	26.9	30.4	32.2	32.6	27.5	26.5	27.5	29.3	28.8	29.1	31.1	30.8
31.4	30.3	30.7	27.4	31.9	33.6	35.5	29.1	28.4	29.9	30.6	31	31.2	32.3	32.4
29.1	28.2	28.4	26.1	30.1	32.4	31.9	27.5	25.8	27.4	28.9	27.4	28.5	30.5	31
29.8	27.6	28.7	26.5	29.6	31.8	33.6	27.7	26.5	27.9	28.5	29	28.9	30.9	30.3
29.4	28.9	29.5	26.5	31	31.8	34.4	27.8	27.8	29	29.5	29.9	29.7	30.7	30.9
29.4	28.6	28.5	25.7	30.4	32.6	33.9	27.8	26.3	28	28.7	28.6	29.4	30.6	30.2
29.3	28.7	28.7	26.2	30.8	32.7	33.1	27.8	26.6	27.3	28.9	28.4	28.5	30.7	30.7
29.1	27.9	28.9	25.8	30.4	31.6	32.8	27.5	26.2	27.7	28.2	28.7	28.8	30.3	30.2
28.5	27.3	27.9	25.3	29.3	30.3	31.7	26.5	25.1	26.7	27.9	27.9	27.9	29.3	29.2
31.5	30.8	30.5	28.2	32.2	33.2	35	28.6	28.8	30.5	31.7	31.8	31.5	30.7	31.2
31.1	29.5	30.1	28.4	32.5	32.6	35.7	28.9	28.6	29.8	29.9	31.2	30.5	31	31.6
28.3	27.4	28.1	25	29.8	29.9	31.7	26.5	24.7	26.9	28.2	27.7	27.7	29	29.1
29.8	28.9	28.6	26	30.7	31.8	35.5	27.9	27.4	28.8	29.4	29.3	29.3	30.3	30.3
30.3	29.7	29.5	26.3	30.9	32.8	34.5	28.2	27.7	29	29.4	30	30.1	32.4	31.7
28.8	27	27.5	24.9	29.4	30.8	33.7	27.1	25.7	27.4	28.2	28.7	28.5	30	30
28.7	27.3	27.6	24.8	29.2	31	34.1	26.5	26	27.4	28	28.3	28.8	29.8	29.8
36.8	33.2	35.5	33.4	35.4	40.5	38.8	35.5	31.9	34.7	34.6	35.4	34.6	35.3	34
36	34.4	34.7	33.9	34.8	38.7	37.8	34.4	33	34.8	35.2	34.9	34.4	35.4	34.1
36.4	33.2	34.6	31.4	34.3	37.6	37.9	33.1	32	33.7	33.4	34.3	32.5	34.1	32.7
34.6	32.9	34	30.5	32.1	34.9	36.6	32.5	31.4	32.6	32.9	33.3	31.1	32.7	32.7
35.2	31.3	32.8	31.1	32.7	38.3	37.6	31.1	30.8	33.2	32.8	34.3	32.7	32.9	31.9
34.3	33.6	34.9	30.7	33.2	34.9	36.6	32.3	32	32.8	33	33.7	32.4	34	33.5
36.8	32.9	34.7	35.3	35.5	40.4	38.9	34.2	31.6	35.1	35.5	35.7	34.6	34.4	35.2
35.2	33.4	34.5	32.3	33.9	37.5	36.8	33	31.9	32.7	33.8	33.9	33	33.6	32.9
34.5	33.4	33.1	33.7	33.4	38.3	36.5	33.7	32.2	34	35.1	34.7	34.2	33.7	35.2
36.4	32	33.8	31.8	34.4	38.4	38.8	32.9	31.2	33.8	33.4	34.4	32.5	33.8	32.4
33.1	31.2	30.2	28.2	31	33.9	36.6	30.7	29.2	30.2	30.8	32.2	32.8	31.4	31.4
35.7	33.1	33.3	31.1	32.8	38.2	39.7	31.6	33.8	34.5	34.2	36.4	34.4	34.3	33
35	32.6	33	30.7	33	35.5	38.1	31.9	32.2	32.6	32.6	35.2	32.7	33.3	32.7
32.4	30.2	31.2	28.5	31.2	34.8	36.5	30.6	29	30.3	31	31.5	32.3	30.7	31
35.5	32.9	33.9	30.9	33.4	35.5	36.6	32.2	32.1	32.8	33.3	33.9	30.8	33.2	32.1
36	32.3	32.8	31.4	33.4	38.3	39	32	32.1	34.2	33.8	35.5	34.3	33.6	32.9
33.8	31.1	31.9	30.2	31.8	36.8	36.7	30.7	29.4	32.7	32.2	32.3	33.1	31.3	32.9

33.3	31.6	31.2	28.9	31.9	34.4	37.1	30.9	29.5	30.5	31.4	32.1	32.7	31.6	31.6
34.1	31.3	31	29	32	35.8	37	31.2	30.8	31.9	31.9	33	32.9	31.8	31.2
35.9	31.8	32.9	32.8	32.8	38.9	38.9	32.3	31.2	32.7	33.8	34.7	34	32.4	33
34.1	31	31.8	31.7	32.2	37.7	37.9	32.2	30.2	32	32.3	32.8	33.4	31.5	32.6
34.9	30.3	32.3	32.3	32.6	37.3	35.2	31.7	29.1	32.7	33.3	32.9	30.9	32.1	32.1
36.5	33	33.7	31.7	33.2	39.5	38.9	32	31.7	33.3	34.5	34.3	33.7	33.2	32.6
33.2	30.4	30.8	28.5	31.5	35.4	36.4	30.5	29.9	30.6	31.6	31.8	32.5	31.2	31
36.5	32.3	33.7	33.4	33.5	39.2	38.2	32.5	31.2	33.9	34.3	34.9	33.2	34.3	33.6
34.7	30.8	32	30.1	31.2	38	37.1	31.6	30.8	32.3	33	33.6	33.7	32.8	33.4
32.9	30.8	31.9	30.3	31.8	37.3	36.4	32.3	29.7	30.8	32.5	32.5	32.8	31.4	33
33.1	30.6	30.5	28.2	30.9	36.3	36.4	30.2	29.7	30.9	31.6	31.8	33.7	31	30.9
36	31.5	33.2	32.5	33.2	39.7	37.9	32.2	30.9	32.9	34.6	34.5	33	33.3	32.9
32.8	31.3	30.4	28	31	33.5	36	30.4	29.4	30.5	30.6	32.5	31.8	31.5	30.9
33.4	30.8	31	28.8	31.5	35.9	37.5	30.5	29.9	31	31.3	31.7	32.8	30.9	31.5
31.3	29.8	30.3	28.1	31.6	32.4	33	30.5	29.6	29.4	31	30.4	30.9	32.2	33.5
30.6	29.9	30	26.6	30.9	32.4	33.6	28.9	28.6	29.5	30.7	29.9	30.2	31.6	32.2
30.7	29.9	30.3	27	31.5	32.3	33.6	28.8	28.7	29.6	30.8	30.5	30.8	31.5	32.1
31.3	30.7	30.8	27.4	32.2	32.8	34.8	29.1	29	29.9	30.7	30.7	30.6	31.9	32.3
29.4	28	28.6	26	30	29.7	30.7	27	26.7	28.3	29.6	28.7	29	28.6	30
30.1	29.6	30.4	28.2	32.1	33	36	28.5	28.2	30.4	30.8	31.7	30.9	30.7	30.9
31.5	30.6	30.8	29	32.2	32.7	33.3	30	29.5	30.9	32.5	31.5	31.7	31.4	32.7
30.2	29.2	29.8	27	30.8	31.2	32.1	28.5	28.1	28.7	30.1	29.8	30.3	30.6	31.7
31.9	30.9	30.9	28.4	32.2	33.3	34.1	29.4	29.8	30.8	32	31.5	31.4	31.6	32.8
30.4	29.2	29.7	26.8	31.1	30.5	32.3	28	28.1	28.9	30.4	30	30	30.3	31.5
32.9	31.9	32	28.7	33.2	33.6	35.1	29.9	30.6	32	32.7	32.5	32.5	32.1	33.1
31.4	31.8	31.4	27.5	32.1	32.4	33.7	28.3	28.9	30.4	30.9	30.7	30.5	29.8	31.8
30	29.3	30.5	26.3	30.7	31.1	34	27.6	28.9	28.5	30.1	30.1	29.7	29.8	30.6
32.6	32.8	33.2	33.7	33.4	33.4	33.2	31.2	32.8	33.9	34	32.9	33	32.4	32.6
33.4	31.7	33.9	33.9	33.3	33.5	33.3	32.5	32.7	35.2	34.2	33.3	33.9	33	32.9
31.6	31.2	32.5	32.5	32.2	32	31.7	31.1	31.2	34.2	33.2	32.8	33.5	32.5	32.5
33.3	31.9	33.4	33.3	33.1	32.9	32.5	32	32.4	34.6	33.6	33	33.9	32.6	32.4
32.9	31.8	33.3	33.7	33.2	33	32.4	32.4	32.4	34.7	34	33.1	34	32.1	32.6
32.5	31.6	33	32.7	32.8	33	32.4	31.3	31.8	34.3	33.4	31.8	33.1	32.3	32.1
32.1	31	32.6	32.3	32.1	32.1	31.6	30.9	31.2	34.2	33	32.6	33.4	32.3	32
33.2	31.3	33.7	34.1	33	32.8	32.7	32	31.6	33.7	33	32.4	33.3	31.9	32.5
34.8	31.1	35.8	33.8	34.2	34.2	34.3	32.7	31.8	34.6	34.2	33.7	35	33.6	33.7
34.2	31.9	35.9	33.7	35.3	35.5	33.8	33.1	31.9	34.7	35.3	33.9	34.4	33.4	33.9
35.7	31.1	35.2	33.8	34.1	37.6	34.6	33.2	31.3	34.7	35	33.6	35.1	32.5	33.1
31.8	30.9	32.2	31.8	32.3	31.9	31.5	30.9	31.2	32.9	31.8	31.7	32.4	31.6	31.4
31.4	30	33.3	33.8	32.5	32.8	31.7	32.7	32.1	34.1	32.9	32	32.8	31.6	31.9
33.1	31.5	34.7	34.5	33.4	34.4	32.7	33	31.9	33.9	33.5	33.1	34.7	32.2	32.5
33.5	30.5	33.9	31.6	32.7	32.6	32.9	32.2	30.7	33.8	33.5	32.6	33.6	32.3	32.8
34.5	32.2	35.1	33.6	34.7	35	34.7	33.2	32.1	34.9	34.7	33.9	34.3	33.6	34.1
33.3	31.9	34.3	33.7	33.5	33.9	33.3	32.2	32.2	34.7	34.1	32.8	33.9	33.1	32.9
32.2	31.2	34.3	34.1	33.4	33.7	32.2	31.9	31.7	34	32.7	32.6	33.1	30.9	32
29.6	28.6	29.7	26.7	30.7	30.3	30	30	28.2	29	30	27.4	29.7	30.2	30
28.9	28	28.7	25.4	30.8	29.7	29.3	29.5	28.1	28.4	30.2	27.2	29.7	30	29.4

29.3	27.5	28.8	25.9	30	29.3	29.8	29.5	28.1	28	30.4	27.5	30.1	30	29.9
29.3	27	28.6	25.9	29.9	28.9	29.9	29.5	27.9	28.1	29.6	27.2	30	29.9	29.9
30	27.8	29.2	25.4	31.1	30.5	30.3	29.4	28.8	28.6	30.2	27.3	30.1	29.8	29.7
27	25.9	27.1	24	28.6	27.1	27.5	27.2	26	26.7	28	25	27.5	27.5	27.4
29.1	27.8	28.8	25.2	30.6	29.9	29.6	29.8	28.4	28.8	30.1	27.5	29.9	30	29.9
29.3	28.3	29	25.9	30	29.9	29.8	29.9	28.3	28.6	29.9	27.2	29.8	30.3	29.8
28.6	27.4	28.5	25.2	30.3	28.7	29	28.5	27.4	27.9	29.6	26.5	28.9	29	28.3
29.7	28	29.4	25.6	30.9	29.9	29.3	29.9	28.3	28.5	31.2	27.6	30.4	29.9	29.8
30.1	28.2	29.8	26.3	31	29.9	29.4	29.9	28.5	28.7	30.5	27.5	30	30.2	29.9
30.3	29.1	30.1	26.5	31.6	30.5	30.3	30.5	29	29.1	30.7	27.9	30.3	30.7	30.4
31	30.2	30.2	28.6	31.4	32.7	31.8	30.2	29.2	29.3	31	30.5	30.4	31.9	32.2
31	30.4	31.2	29	32.2	32.9	32.6	31.1	29.9	30	31.5	30.8	30.9	32.3	32.7
32.3	31.1	31.3	29.4	33.3	33.5	32.8	31	29.9	30	32	31.4	31.2	32.5	33.4
31.9	30.6	30.7	28.6	32.7	33	32.1	30.6	29.6	29.4	31.8	30.7	30.7	32	33
30.9	31.3	31.1	29.7	33.1	32	33.1	30.9	29.5	29.9	31.7	31.6	31	32.9	33.1
30.9	30.8	30.7	29.3	33.2	32.4	32.9	30.7	29.6	30	32	31.4	30.6	32.3	32.7
31.5	30.8	31.1	29.7	32.8	33.6	32.9	31.2	30	30.4	32	31.3	30.8	32.6	33.2
31.5	31.4	31.2	29.9	33.3	33.2	33.4	31.1	29.9	30.2	32.2	31.8	31	33.1	33.5
31.8	30	30.1	27.8	32.6	33.5	32.5	31.3	29.6	30.5	31.4	30.8	30.2	31.4	33.5
34.1	34.1	32.9	31.5	35	35.5	34.9	32.6	31.7	31.7	33.9	33	32.4	34.4	35.1
31.2	30.9	30.5	28.9	32.4	32.7	32.3	30.4	29.4	29.4	31.8	31	30.6	32	32.7
31.7	30.8	31	29.5	32.7	32.7	33.1	30.7	29.7	29.9	32.4	31.7	31.1	32.7	33
32.1	31.2	31.1	29.7	33.1	33.3	33.3	31	30.1	30.3	31.9	31.4	31.2	32.7	32.8
30.9	30.6	30.3	29	32.5	32.2	32.5	30.1	29.3	29.5	31.5	30.8	29.9	32.2	32.7
32.2	32.3	32.3	30.5	32.8	33.4	33.8	31.8	31.6	31.3	32.9	33	31.2	33.2	34
27.5	25.5	27.1	24.1	28.9	29.2	29.4	27.2	26.2	26.8	28.1	27	26.8	28.1	29.6
27.7	25.2	26.7	23	27.6	27.3	27.7	27.1	26.9	25.3	26.3	25.5	28.1	28.1	27.1
27.6	25.8	27.9	23	27.8	27.8	27.8	27.9	28.2	25.4	26.7	25.7	28.4	27.2	27.7
27	25	26.6	23	27.2	26.5	27.1	26.5	26.6	24.6	25.7	25.7	28.3	27.5	27.1
28.1	25.6	27.2	23.3	28.1	27.9	27.9	27.4	26.5	25.7	27.3	26.3	28.1	28.5	27.6
26.1	23.4	25.2	22.1	25.7	25.5	25.9	25.7	24.9	24.3	24.7	24.4	26.3	26.3	25.7
26.4	24.7	26.3	23	26.8	26.2	26.5	26.2	26	24.3	25	25	28.5	27.1	26.8
26.9	24.5	26.3	22.9	27.4	27.7	26.9	26.1	25.3	25.2	26.8	25.3	26.3	27.7	27.1
24.8	22	22.5	20.2	24	24.3	24.8	24.5	24.4	22.2	23.5	23.1	25.2	24.5	24
27.6	25	26.3	22.6	27.4	27.2	27.3	26.5	25.6	24.9	26.2	25.6	27.1	28.1	26.9
27.8	25.2	27.4	22.7	27	27.7	28	27.2	26.6	25.6	27.2	26	27	27.9	27.2
25.9	24.3	25.9	21.5	26.2	26.7	26.8	26.9	24.9	24.4	25.5	24.6	26.2	26.7	25.5
28.3	26.4	27.8	23.5	28.5	27.4	28.1	27.7	27.3	25.7	27	26.2	28.4	27.9	28.1
29.5	27.4	28.5	24.9	29.7	31.2	31.5	28.5	25.8	28	29.2	28.3	30.2	29.4	31.2
25.7	24.3	23	19.8	25.3	26	26.5	24.2	21.8	24.2	24.4	23.9	26	25.3	25.5
27.7	26.3	26.6	24	28.3	28.6	30.8	27	24.6	26.2	27.5	26.2	28.8	28.4	29.5
28.3	25.3	24.7	21.1	26	27.7	28	25.5	23.3	25.8	26.5	25	26.9	27.4	27.1
28.9	28.8	28.5	25.8	30.2	31.3	32.5	28	25.7	26.8	29.2	28.3	29.1	30	30.5
28.9	27.2	27.7	25.3	29.4	30.9	29.5	26.8	25.7	27	28.7	27.5	28.2	29.6	29.9
29.8	27.4	27.2	23.7	27.3	29	29.9	26.6	24.6	27	27.5	26.8	28.1	28.2	28.2
27.7	25.9	24.8	21	25.6	27	27.5	24.2	22.6	24.8	24.9	24.3	25.3	25.7	26.2
29	27.5	27.2	23.5	28.8	29.7	30.9	26.9	24.8	26.8	27.9	27.2	28.7	28.7	30.3

25.5	23.1	23.5	21	24.2	25.5	26.5	23.1	21.2	23.2	24	23.2	24.1	24.3	25.2
28.5	26.2	25.4	22.4	26.6	26.9	27.5	25.5	24	26	26.3	24.3	27.2	25.7	26.9
28.9	27.6	28	25	29.4	31.9	32.5	27.3	25.6	27.2	29.8	28.5	29.2	29.6	30
28.1	27	27.2	25	29.4	30.4	31.1	27.9	25.7	27.1	29.2	28	29.3	29.3	30.2
25.7	24	24.9	22.6	27.1	26.5	29.1	25.5	22.1	22.1	25.6	24.7	25.9	25.8	27
29	28.6	28.3	25.8	30.5	31.9	33.1	27.3	26.3	26.8	29.3	28.4	29.3	30	30.7
28.5	26.1	26.4	24	27.5	29.1	28.7	26.8	24.4	26.4	27.8	25.6	28	27.4	29
27.9	26	25.2	21.9	25.9	26.9	27.9	25.2	22.2	25.2	25.9	25.4	26	26.6	26.4
26.4	24.4	23.1	20.1	24.9	25.8	25.9	23.5	21.4	24	24	23.2	24.4	25	25.5
29.1	28.9	28	25.4	29.9	31	32.9	27.9	25.8	27.2	29.2	28.6	29	29.8	30.5
28.1	27.7	27.2	24.7	29.1	30.4	32	26.9	24.9	26.5	28.5	27.9	28.8	29.3	29.7
27.9	27.7	26.9	24.3	28.7	30	32.6	26.9	24.6	25.9	27.6	27.6	28.2	28.9	29.2
28.5	27.8	28.1	24.8	29.1	30.6	31	28	25.2	26.9	28.8	27.8	28.6	29.2	30.1
27.2	25.7	25.9	23.4	27.4	28.6	29	25.5	23.4	25.5	27	25.7	26.2	27	28.1
25.8	23.4	22.1	19	24.3	25.5	26	23.2	21.2	22.8	23.6	21.9	24.4	24.5	24.4
26.6	22.8	22.3	20.7	24	23.7	25.9	20.6	21.9	23.5	23.1	21.6	22.6	23.7	23.9
29.3	26.3	25	21.9	26	27.2	27.9	24.8	24.2	25.9	25	24.8	25.8	26.5	27.4
29.9	27.8	25.9	23	26.9	28.4	29.4	25.9	25.6	27.5	26.4	27.5	27.8	28.1	28.9
29.2	27.6	25.1	23.1	26.8	28	29.6	25.6	25	26.7	26	26.7	27.1	27	27.8
29.1	27.5	25.8	22.7	26	27.9	28.8	26.3	24.8	26.9	25.7	26.2	26.9	27.2	27.4
28.9	26.5	24.7	21.1	25.5	27.1	28.6	26.4	24.1	26.2	25	25.5	26.3	27.1	26.8
29.3	27	25.7	22.4	26.1	28.3	28.1	25.4	24.5	27.4	25.8	26.1	26.4	27.2	28.1
30.1	27.6	25.7	23.1	26.5	28.3	29.7	26.1	24.7	26.7	25.7	26.8	26.3	26.6	27.5
32.3	30.6	29.7	26.5	30.1	31.2	31.9	28.9	27.8	28.9	29.1	29.1	29.2	29.4	30.6
32.1	29.8	29.6	26.4	29.7	31	32.2	28.9	27	28	28.9	29.3	29	28.8	29.6
30	28	26.7	23.8	27.1	28.7	29.6	26.6	25.4	27.5	26.1	26.8	26.8	27.5	28.7
31	28.5	26	23.6	27.9	28.9	30.8	27	26.1	27.9	27.4	27.2	28	28.3	30.2
29.7	26.7	25.3	22.4	26.1	27.5	28.9	25.4	24.4	26.4	25.4	25.9	25.4	26.3	26.6
28.4	26.2	24.1	21.6	25.3	26.8	28.1	24.8	23.8	25.7	24.8	25.4	25.7	26	26.9
30.6	29.3	27.5	23.6	26.6	28.4	29.6	26.5	25.3	27	26.4	27.2	26.8	27.2	27.9
30.1	27.3	25.6	23.6	26.4	27.8	29.6	26.5	24.9	26.5	25.9	27	26.4	26.7	27.3
29.4	28.4	27.7	24.9	27.8	29	29.9	27.3	25.6	27.2	26.7	27.8	26.7	27.3	28.4
31.5	30.6	29.9	26.8	29.4	30.6	32	28.9	27.6	29.1	28.7	30	28.9	29.5	30.1
31.9	29.6	27.4	25.9	28.7	29.4	31.6	28.6	26.8	28	27	29	28.8	29.2	29.3
30.5	30.3	29	25.9	29	30.9	31.9	29	25.7	27.6	27.8	28.9	27.7	27.9	29.1
30.2	28.3	28.2	24.2	28.3	29.9	31	27.5	25.8	27.2	28.1	28.1	27.7	27.8	29.3
30.4	28.9	28.9	24.4	27.7	29	29.5	26.6	25	27.1	26.9	26.7	27.1	27.7	28.8
30.7	28.8	26.4	24.5	27.4	28.6	30.6	27.2	25.5	27.6	26.6	27.8	27.5	27.5	28.6
28.9	27	27.1	23.5	26.9	28.2	30.4	26.4	24.2	25.5	26.4	26.7	26.5	27.4	28.1
29.6	27.9	26.9	23	25.5	27.9	29.3	26.2	25.3	26.7	26.6	26.7	26.4	26.5	27.5
30.5	28.4	28.9	26	28.9	30.5	31.4	28.4	26.2	27.6	27.8	28.9	27.7	28.2	29.3
29.6	28.2	26.4	24.7	28.1	28.9	30.6	27.4	25.7	26.7	27.2	27.9	27.4	27.9	28.6
30.7	28.8	28.3	25.1	28.4	29.9	31.7	28.1	26	27.3	27.2	28.2	27.6	28.5	29.4
31.1	30.1	28.8	25.8	29.3	30.6	32.1	28.2	26.4	27.5	27.6	28.2	27.6	27.3	29.3
29.3	28.6	28	24.4	27.7	29.4	30.7	27.2	25.2	26.5	26.4	27.4	26.5	26.8	28
30	29.8	28.3	25.4	28	29.8	31.2	27.4	26.2	27.4	27.4	28.4	27.1	27.4	28.6
30.1	28.9	28.5	25	28.2	30.4	31.7	28.4	26.6	28.3	28.1	28.8	27.5	27.6	28.7

31.2	30.3	28.8	25.3	28.3	30.3	32	27.9	25.5	27.9	27.7	29.1	27.8	28.3	29
32.2	30.3	29.2	28.3	30.8	33.1	35.4	30.2	28.5	29.8	30.9	31.4	31.4	30.7	30.6
31.9	31.2	30.7	28	32	34.1	36.3	29.5	29	30.9	30.8	33.2	30.9	30.8	32
32.5	30.6	29.8	27.4	30.9	33.8	36.4	29.5	29.3	30.2	31.4	32	31.8	31	31
30.9	28.9	28.7	26.5	29.8	32	34.8	28.2	26.8	27.9	29.3	30.8	30.1	30.2	30
33	30.2	30.1	28.4	31.1	33.8	36	29.7	28.9	29.9	31	31.8	32.2	31.5	31.6
32.6	30.6	32.1	30	32.1	35.2	36.3	31.2	29.5	31.4	31.2	31.2	32.4	30.5	32.7
32.8	30.9	29.4	28.4	32.2	35.6	37	29.9	29.1	30.5	30.6	31.6	32	31	32
31.4	29.2	29.9	27.6	32.1	33.6	35.9	29	27.6	29.2	30.8	31.9	30.9	29.1	31.2
31.3	31.2	30.7	29.9	32.5	33.7	35.6	29.4	29.1	30.7	31.5	31.8	32	31	32.5
33.6	30.5	30.9	28.8	32.5	34.7	37.1	30	29.9	30.7	31.5	32.3	32.8	31.1	32.6
32.2	31.1	34.8	29.8	33.2	33.8	33.7	30.4	29.6	32.5	32.5	32.7	32.7	31.2	33
31.7	28.8	28.9	27.7	30	32.9	35	28.9	27.5	28.9	29.7	30.2	30.7	29.4	29.6
31.9	30.1	30.2	28	31	33.8	36.4	30.1	28.6	29.9	30.6	32	31.8	31.4	31.5
31.4	28.9	28.8	26	29.2	32.1	34.6	29	27.2	28.5	29.5	31	30.5	30.5	30.1
32.5	30.4	29.1	27.1	30.1	32.9	35.8	29.4	27.6	28.8	29.5	31.1	32	30.7	30.4
32.2	29.8	31	28.5	32.2	36.2	36.3	30.5	29.6	31.2	31.9	32.9	33.1	31	32.5
33.2	30.7	31.8	30.4	32.2	36.9	37	30.9	30.2	32.1	33	32.8	33.8	31.2	32.5
33	31.2	31.3	28.9	32	35.7	37.5	30.6	29.3	31.3	31.5	32.2	31.8	31	31.9
32.5	31	32.6	29.9	32.7	33.6	36	30.4	29.5	31.9	31.9	32.5	33.4	31.6	33.5
33.7	30.6	30.9	28.9	31.6	35.7	37.3	31	29.1	30.9	31.5	32.8	31.9	31.1	32.3
31.3	31.3	28.8	27	29.7	32.6	35.2	29.6	27.5	28.4	29	30.5	31.5	30.8	30.6
33	30.2	30.2	28.2	31.4	34.2	36.5	30.2	29.3	30.3	30.9	31.9	31.3	31.5	32.2
32.5	31	29.2	26.9	30.1	32	35.3	28.9	26.8	28.1	29.2	30.7	30.9	30.3	30.2
32.1	29.4	30.3	28.1	31	35	36.2	30	28.5	30	30.5	32.2	31.8	30.4	32.3
32.6	29.7	29.5	28.2	31.2	33.9	36.4	30.3	28.6	30.2	30.9	31.7	31.9	30.8	31.1
32.5	30.2	29.7	27.8	30.9	33.5	35.8	30	27.7	29.7	30.2	31.2	31.3	30.8	30.7
33.1	31.9	33.6	33.4	33.3	33	32.5	30.8	31.7	34.7	33.1	31.4	33	32.3	31.9
35.3	31.7	34.5	32.1	34	34.7	34.2	31.8	31.8	34.3	34.2	33.6	33.1	32.8	33.6
36	32.9	35.5	35.1	35.3	34.8	34.7	33	34.3	35.3	34.9	33.4	34.7	34	33.6
34.8	32.5	35.4	33.3	33.5	34.3	33.3	31.5	31.2	35.3	34.3	32.9	33.8	33.3	33.2
35.2	31.9	33.9	32.2	34.8	34.4	33.9	31.7	31.2	34	34.7	33.6	33.6	33.5	33.8
34.5	30.5	33.5	31.3	33.1	34	33.1	30.9	30.5	32.6	34	31.9	32.4	31.9	32.6
36.6	33	35.3	33.4	36.3	35.2	34.3	31.8	31.9	35.9	35.9	35	34.5	34.4	34.6
33.5	32.2	33.4	30.1	33.1	33.6	34	29.8	29.2	32.2	32.5	32.1	31.8	31.1	32.6
34.1	31.9	34	30.9	33.7	34.4	34.3	31.2	29.9	33	33.5	33.1	33.3	31.9	33.4
32	31.6	32.5	31.9	31.7	32	31.5	30.2	31.2	32.6	32.2	31.2	31.8	31.8	31.3
34.2	31	35	29.7	32.7	33.2	32.9	29.5	29	31.5	32	32.2	31.6	31.1	31.6
34.1	30.4	33.5	32	32.7	33.2	33.1	30.5	30.4	33.3	33.2	31.8	32.8	32.2	32
36.3	34.3	37.2	36.3	36.1	36.2	35.8	33.9	34.3	37.1	36.2	34.8	36	35.2	35.1
34.1	32.1	34.4	34.5	33.8	33.9	33.9	32	31.9	34.6	34	32.9	33.6	33	33.5
34.8	32	34.9	33.3	34.3	33.8	33.3	32.2	31.8	34.7	34.6	33.3	33.5	32.8	33.3
33.9	32.2	34.1	34.4	34.3	34.1	33.5	33.2	32.6	35	34.4	33.5	34.3	33.2	33.3
34.8	31.4	34.2	33	34.2	33.8	33	31.5	31.3	33.9	34.5	33	33.2	32.7	32.2
33.3	31.1	34.1	32.7	33.2	33.8	32.5	31.4	30.5	33.2	33	31.8	32.7	31.8	31.9
32.9	30.9	34.1	32.7	32.2	33.2	33.2	31.2	31	33.8	32.9	31.2	32.6	32	31.9
33.4	30.7	33.8	31.3	34.3	35.1	33	31.2	30.4	33.2	33.3	32.7	33.2	31.8	32.9

33.8	31.9	34.4	33.8	33.2	33.6	33	31.1	32.2	34.6	33.4	32.3	33.4	32.5	32.2
34.2	31	34.8	31.2	33.6	34.9	34.3	31.2	30.1	33.3	33.5	32.7	33.3	31.8	32.9
36.4	33.1	35.7	34	36.2	35.3	34.7	33.4	32.4	35.3	35.3	34.5	34.8	34.1	34.8
35.8	31.9	34.2	32.1	34.5	34.4	34.5	31.5	30.4	33	34.5	33.1	33	32.4	33.3
34.9	31.7	34.3	31.4	34.2	35	34.8	32.3	30.8	33.3	34.2	33.9	34.2	32.8	34.1
31.8	30.9	31.6	31.8	31.8	31.9	31	30.3	30.8	33.2	31.4	30.9	31.9	31.6	30.8
35.2	32.6	35	34.3	35.5	33.7	34	32	32.5	35.2	34.9	32.9	34	33.4	33.3
34.5	31.4	33.6	30.6	33.7	33.6	33.5	30.3	29.6	32.4	33.2	32.5	32.4	31.9	32.5
34.8	31.5	33.9	30.8	33.5	34	33.8	31.7	30.1	33.5	33.8	33	33.6	32.3	33.3
33.6	30.4	33.7	31.5	34.3	33.8	32.7	31	29.8	33.2	32.7	31.8	33	31.6	32.5
34.7	31.8	34.6	31.9	34.7	35.3	34.8	32.9	31.9	34.4	34	33.8	34.1	32.8	32.1
34.3	31.6	33.5	29.9	33.2	33.9	33.8	30.6	29.3	32.6	32.6	32.1	32.5	31.1	32.7
31.7	33	29	28.7	27.4	27.7	30.8	32.1	30.6	28.4	28.1	32.7	29.6	27.1	27.5
29.7	31.4	28.8	26.7	26.4	27.8	29.8	29.6	28.9	26.4	28.4	30.9	29.5	27.5	28.5
33.8	36.1	30.9	31.9	28.7	28.2	30.8	33.2	32.3	31.8	29	34	31.8	29.5	29.9
34.3	34.4	32	29.1	29.2	31.7	34.1	31.3	31.1	30.6	31.4	34.4	30.1	29.2	30.9
31.6	33.7	29.1	29.1	27.4	28.2	30.5	31.8	31.1	28.6	27.7	32.9	29.8	27.6	28
30.5	32.7	26.5	25.4	24.2	26.6	28.8	27.6	28.5	26.9	26.1	31.3	28.1	26	26.9
29.3	30.6	26	25.2	25.6	26.1	28.7	28.7	28.2	25.8	25.8	29.3	28.1	26	25.8
28.8	30.4	27.1	25.1	25.4	26.5	28	28.7	27.8	25	25.6	29.5	27.4	25.7	25.9
28.3	32.8	28.2	27.8	25.8	26.1	28.1	29.6	28.8	28.7	25.6	31.4	28.8	26.1	26.8
31.5	31.8	28.6	27.2	27	29.1	31.6	29	29.2	27.2	27.9	32.1	27.6	26.6	27.6
28.7	30.6	27.5	24.8	25.4	26.4	28	28.3	27.7	24.9	26.8	28.7	28	26.9	26.2
31	32.4	29.2	27.6	26.1	28.6	31.1	30.1	29.3	27.1	29.1	31.2	28.9	26.7	27.3
28.6	29.7	26.4	24.1	24.3	26.3	28.6	28.2	27.2	25.1	27.2	29.6	27.3	23.8	25.6
32.2	34.7	29.8	30.5	29.3	28.6	31.8	32.6	31.7	29.8	28.2	33.7	30.7	28.5	29.1
34.2	33.6	30.1	28.4	28.7	30.1	33.2	28.5	30.6	30.1	27.9	34	31.5	29.6	30.4
32.8	34.1	29.7	28.4	27.8	29.6	32.1	30.9	30.3	29.5	30.2	34.7	29.7	28.2	29
30.6	33.8	28.3	27.7	26	26.5	28.8	30.7	29.8	28.6	26.9	32.3	29	27.2	27.4
33.4	34.4	30.3	29.5	29	30.4	32.9	29.6	31.3	29.2	29	32.6	28.7	28.5	29.9
33.8	35.3	30.9	29.1	28.1	30.2	32	31.9	31.6	29.4	30.2	34.5	30.6	28.8	29.6
29.8	31.6	27.5	26	25.8	27.7	29.2	28.2	28.2	25.9	27	31.3	26.9	25.3	26.7
30	30.2	27.2	25.3	25.9	27.4	29.8	26.6	28.1	25.5	27.5	28.8	26.4	24.6	27.7
34.3	33.5	30.8	29	30.5	30.3	34.2	30.4	30.6	30.2	31.1	34.4	28.5	27.6	30.6
29.3	31.9	27	24.6	25.1	27.8	28.9	28.4	27.2	25.7	26.5	30.3	26.6	25.3	25.9
30.3	32	28.8	26.6	27	28.5	30.6	29.8	28.9	26.6	28.7	31.2	29.1	27.2	28.2
28.1	30.2	25.9	24.3	23.3	24.6	27.7	27	26	25.1	25.2	29.8	26	24.2	25.4
27.5	29.5	26.2	23.5	24.1	25.2	25.9	26.7	26	22.9	25.2	27.1	26.4	24.2	24.1
33.3	33.1	30	28.2	26.9	29.5	31.9	28.4	30.5	28.8	28.2	34.1	30.2	28.4	28.5
32.8	34.5	29.3	27.9	27	29.8	32.5	30.7	30.1	28.2	28.9	33.7	28.6	28.2	28.2
32.7	33.5	29.8	28.4	29.4	30.8	32.8	31	30.9	29.3	29.9	32.5	28.8	27.3	29.2
29.8	31.9	26.9	25.5	24.9	26.5	28.8	26.8	27.9	26.3	25.8	30.7	27.2	25	26.3
33.9	33.9	30.5	31.4	28.5	31.1	34	28.6	30	29.2	28.3	34.2	30.8	29.3	29.8
29.5	29.5	26.6	24.3	25.9	27.3	29.3	27.8	27.9	25.4	26.8	28.4	27.2	24.7	26.6
34.2	35.5	31.7	29.6	29.1	31.3	35.2	31.1	31.2	31.5	31.9	35.5	30.9	29.6	31.3
34	33.6	30.7	29.2	28.3	30.6	35.2	29.3	31.4	30.5	29.7	35.5	31	29.3	30.2
33.1	33.7	29.5	28.4	27.5	29.7	32.7	27.5	28.9	28.9	27.4	33.2	30	28.1	28.6

30.4	31.1	27.2	25.7	25.4	27.3	29.2	28.4	27.8	26.8	26.6	31.1	27.7	25.3	26.8
32.9	32.1	28.8	26.6	27.3	28.7	32	28.5	28.2	29.6	30.4	33.4	28.9	26.8	30.1
32.3	31.9	29.3	25.7	27.1	28.5	32	25.9	28.5	29	28.6	33.2	28.3	27.3	28.7
32.9	32.8	29	28	26.5	28.8	30.7	27.3	28.9	28.2	26.8	32	28.6	27.3	27.8
33.7	34.1	32	29.1	28.6	30.7	34.4	29.9	31.1	31.4	31.6	35	30.9	28.9	30.8
33.3	33.2	29.4	26.4	28.8	29.6	32.5	28.3	28.1	29.3	29.7	32.9	28.5	26.7	29.4
33.3	33.2	30.7	26.8	27.8	29.7	33	28.7	28.5	29.7	29.4	34	29.4	27.6	30.1
34.4	34.5	32.1	27.9	30.1	30.1	33.6	29.4	31.3	31.4	29.9	34.4	30	29.9	30.7
34.2	34	30.6	27.4	28.2	29.7	33.5	26.7	29.6	29.6	28.6	33.7	30	28.5	28.5
27	25.5	27	24.6	28.4	28	28.2	25.6	25	26.6	28.2	27.6	26.6	27.3	28.1
30.8	29.8	30.6	29.2	31.9	31.7	32	29.8	29.1	30.3	31.3	30.9	29.7	32.2	31.8
32	31.9	32.2	31.3	32.7	33.7	34.2	30.4	30.7	34	34.8	32.7	32.9	34.2	33.6
30.7	30.4	30.9	29.9	29.8	31	31	30.6	30	30.3	30.6	29.8	30	31.1	31.7
31.2	29.8	31	29.6	32.6	32	32.5	29.4	29.4	31.7	32.6	32.1	30.8	31.5	32.1
31.4	32	31.3	30.8	33.4	34.5	33.8	30.9	31.1	32.5	32.6	33	32.3	33.5	34.6
33.6	33.3	33.2	32.7	35	35.5	36.3	32.4	31.7	33.9	35.5	34.6	34.1	35.6	35.3
31.6	31.9	30.9	30.4	33	34.2	34	30	31	31.3	32.8	32.5	31.8	33.5	33.3
30.9	29.2	30.3	29.3	31.6	31.8	32.3	28.9	29.7	31.1	32.3	31	30.8	31.6	32
32.7	32.8	32.2	31.7	33.1	34.3	34.2	29.6	31.5	33.5	33.9	32.9	32.3	34	33.1
30.9	30.8	30.8	29.1	32.1	32.3	32.5	30.3	29.5	30	32.1	30.9	29.5	31.5	32.4
25.6	23	25.6	23.7	26.3	25.2	25.2	22.6	23.1	24.8	26.5	25.7	24.4	25	25.5
30.8	30.7	31.7	30.1	32.5	32.6	33.4	29.6	30.2	31.2	32.5	32.1	30.8	32.1	32.8
32.1	32.8	31.9	31	33.2	34.3	34	30.6	31.1	31.4	32.4	32.6	31.4	32.2	33.3
33.5	32.7	33	32.6	33.4	33.9	34.1	32.2	31.6	33.5	34.5	33.2	32.4	33.6	33.8
31	29.7	31.4	29.2	33	31.6	32.3	29.5	29.1	31.7	32.9	31.8	30.8	31.5	32
30.5	29.4	30.8	28.7	32	32.2	32	29.5	30	30.8	31.6	31.5	30.8	31.5	32
28.4	27.4	27.9	27.7	28.5	29	29	28.2	28.3	28.9	29.9	28.5	27.9	29.2	29.6
31.7	31.7	32.1	30.9	33.2	33.6	33.3	31	30.5	32.3	32.8	32.4	31.4	33.7	33.9
31.7	29.8	31.1	30.1	32	33.2	34.3	29.6	30.7	29.9	33.2	32.3	31	32.8	32
27.7	26.4	27.3	25.6	28.7	28.6	28.7	26.1	25.5	27.5	29.4	28.1	27.5	27.5	28
31.9	31.2	32.3	31.2	33	35.3	34.2	31.3	30.7	32.5	33.8	32.4	30.9	33.3	33.2
31.6	30.3	30.7	29.8	32.4	32.5	33	29.2	29.9	32	33	31.4	30.7	31.6	32.2
31.9	31.5	31.1	31.2	34.1	34.2	33.5	30.2	31.8	32.7	34.1	32.5	32	33.6	32.5
32.3	31.4	31.8	31.5	33.4	34.3	33.7	30.7	31.3	32.6	33.6	33.1	31.8	33.8	33.3
29.2	28.4	30.2	27.6	31.8	30.8	30.5	28.3	27.7	30.2	31.7	29.6	29.3	29.5	30.4
28.8	27.7	28.9	26.8	30.7	29.9	30.1	26.9	26.7	29.5	30.8	29.5	28.2	29.2	29.9
30.8	29.4	31.3	29.5	32.5	31.7	31.8	28.8	28.8	31	32.1	31.2	29.6	31.5	31.9
32.7	32.5	32.6	31.4	33.9	34.6	34.2	31.4	31.7	32.6	33.4	32.8	33.7	33.9	33.7
32.4	31.6	29.8	24.1	27.4	27.4	31.6	27	26.4	28.4	28.4	31.4	27.5	26.8	27.9
33.5	32.8	31.3	26	28.5	29	32.7	28.5	27.7	29.5	29.5	33.2	28.8	27.8	29.5
33.7	32.4	29.3	26.2	28.1	29.2	32.3	28.9	27.2	29.1	30.1	32.8	28.3	26.9	29.3
27.5	26.5	23.9	22.8	25.5	26.3	27.6	25.3	23.6	25.4	24.1	26	25.9	25.9	26.4
32.8	31.7	30.9	25	28.1	28.6	32.1	27.6	27	28.8	29	32.7	28.2	27	28.8
30.5	28.8	26.8	23.9	25.5	26.9	29.5	24.8	24.2	26.9	25.9	28.3	26.5	26	26.8
32.8	33	31	25.4	27.7	28.2	31.8	27.4	27	28.9	28.6	32.8	28.3	27.3	28.6
30.1	28.2	26.3	24.5	27.5	27.8	30.1	27	25.7	27.5	26.5	28.3	27.8	28.2	28.6
31.9	30.6	28.6	25	27.4	28.2	30.8	25.6	26	29.5	27	29.7	28.4	27.3	28.2

32	29.1	27.5	23.2	26.2	27.1	29.6	26	24.9	26.7	26.3	29.7	26.6	26.2	27.4
30.6	29.8	27.8	24.4	27	27.5	29.9	26.2	25.2	27.6	26	29.1	27	26.3	27.4
29.6	27.9	25.5	24	27.1	27.9	28.8	26.4	25.3	27.1	25.7	27.3	27.7	27.4	28
32.8	32.2	30.2	26.1	28	27.9	31.5	27.1	26.5	28.7	28.8	32.1	28.6	28.3	28.6
31.1	29.1	27	25	28.5	28.6	31	27.3	25.9	27.6	27	28.6	28.7	28.7	28.5
33.7	31	29.6	24.8	28.2	28.8	31.3	27.7	26.3	28.3	28.1	31.2	28.2	27.5	28.9
30.7	29.8	27.3	23.2	25.3	28	29	25	26.5	27.9	26.5	30.3	27.5	27.1	27.5
35.7	32.8	30.8	27.5	30.2	31	33.6	29.7	28.4	30.5	29.7	33.5	30.3	29.5	31.3
31.8	29.5	27.9	25.4	28.3	30	31.4	28.5	26.5	28.5	28	28.9	28.1	28.4	29
31.6	29.6	28	25.4	28.4	29.3	31.4	27.5	26.5	28.6	28.1	29.6	27.8	28.1	29.4
31	29.3	27.3	25.5	28.4	29.3	31.6	27.6	26.3	27.9	27.5	28.8	28.3	28.6	29.2
32.9	30	28.5	25	27.6	28.5	31	27.7	25.8	27.8	26.5	30.3	27.3	27.3	28.7
31.7	29.5	27.7	25.7	27.9	28.9	31	28.2	26.2	28.1	27.2	29.7	27.8	28	29.3
30.1	28.9	27.2	25.6	27	26.7	31.1	28	25.7	27.5	26.8	28.2	27	27.7	28.8
31.6	28.7	28.7	24.9	27.6	28.4	29.4	26.6	25.3	27.9	27.1	29.6	26.9	26.7	28.1
34.9	34.3	34	30.1	32.1	33.8	36	31.1	31.3	32.1	32.9	34.3	31.9	31.8	34
32.7	31.8	30.2	26.7	30.5	32.1	36.1	29.6	28.1	29.8	29.7	32.5	29	29.6	30.7
31.5	31.5	32	27.9	29.8	31.3	33.2	29.3	28.8	28.7	30.4	32.2	29.9	29.7	31.3
33.7	31.5	31	28.6	30.8	33.3	36.1	30.5	29.3	31	31	33.8	30.1	30.8	31.7
34.6	33.2	33.9	30.5	32.3	33.5	36	31	31.5	31.4	32	33.4	30.4	31.4	33.1
33.2	31.3	32.6	28.5	29.9	32.7	34.3	29.9	30	29.1	30.4	32.5	29.6	30.5	31.9
33.2	32.1	32.1	28.6	28.7	32.3	35	30.3	30.3	30.1	31.6	33.4	30.3	30.6	32.7
34.2	32	32.4	30.5	31.2	34.5	36.9	31.6	32	31.7	33.5	33.2	33.1	32.1	31.7
32.7	32.5	31	27	28	30.2	32.8	28.9	28.4	28.3	29.4	31.4	28	28.5	30.4
34.7	31.8	33.8	29.6	30.9	33.4	36.1	31.6	32.5	32.3	32.4	34.5	31.5	33.2	33.2
34.2	33.1	31.3	27.5	30.5	32.7	35.6	30.5	30.4	31	31.1	34.3	30	31	31.7
34.6	32.2	33.2	30	32.6	34.4	37.7	32.2	31.3	32.6	32.3	34	31.7	32.6	32.1
32.8	31.4	30.5	28.6	30.5	32.1	35.2	30.4	29.3	31.1	31	32.2	29.8	30.6	31.1
32.9	32	31	27.5	30.2	32.4	36.2	30.3	29.1	29.4	29.6	30.9	30.1	31.2	30.9
32.2	31.5	30.8	28.6	30.5	31.7	34.9	29.9	29	29.6	30.5	31	29.1	30	30.4
32	29.7	29	26.8	29	30.4	34.5	28.7	27.2	29.5	29	31.4	28	28.7	29.4
34	32.2	31.4	28.7	31.5	33.5	35.6	30.9	29.3	30.7	30.8	32.8	30.8	31.7	32.3
32.8	30.4	30.4	27.3	29.5	30.8	35.5	29.2	28.7	30.8	29.1	32.2	29	29.5	30.6
31.9	30.7	31	28	29.8	31.6	34	29.7	28.6	30.1	30.2	31.5	29.5	30.9	30.4
32.7	31	31.9	29.3	30.3	32	34.7	30.1	29	29.4	29	32.6	28.7	30	30.5
33.5	31.6	32	29.2	31.4	33.8	35.8	30.7	30.4	31	31.2	32.9	31	31.6	31
33.4	31	30.8	28.9	30.3	32.5	34.7	29.8	29.4	30	30.6	31.6	30.4	30.7	30.4
32.2	30	28.5	25.7	29.1	31.5	34.5	29	26.6	28.4	28.8	30.5	28.2	29.4	30.4
32.3	31	30.7	28.3	29.8	31.2	35	29.8	28.3	31	30.1	31.3	28.9	30.1	30.7
32.9	31.8	31.3	28.9	30.8	33.2	35.5	29.8	29.7	30.1	29.8	32.6	31.2	31	30.5
33.5	31.5	31	28.2	30.1	33.1	35.5	29.9	29.2	30.2	31.1	32.7	30.2	31	30.2
32.1	30.4	30.7	28.2	30.2	31.9	34.4	29.1	29	30.5	30.1	32.2	29	29.9	30.3
35.8	32.1	33.8	30.6	32.7	33.6	37.8	32.5	32.1	32.9	33	36.1	32.3	31.8	32.4
32.1	30	30.5	28.7	30	32	34.8	29.7	28.8	30.2	29.7	31.6	29.7	30.6	30.7
32.3	31	30	27.5	29.7	31.7	34.5	29.4	27.5	29.2	29.1	30.6	28.8	31.3	30.8
33	32.5	32.1	29.6	31.5	33.4	36	31.2	30.2	31.3	31.2	32.9	31	31.8	31.9
35.4	33.7	34.4	31	33	34.9	37.4	32.7	32.8	33.2	32.3	34.9	31.7	33.3	33.8

31.1	29.9	30	26.6	29.3	30.6	34.4	28.7	27.2	28.8	28.6	31.6	28.4	29.2	29.3
32.1	31.2	29.9	27.5	29.5	31.1	35	29	27	28.6	29.5	31.5	28.9	29.9	30.7
32	30.6	29.8	26.5	29.4	31.4	34.7	29.3	27.9	28.8	29.1	30.7	28.9	30.2	30.3
31.5	30.5	30.1	28.2	30.1	31.7	33.3	29	28.2	29.1	29.6	31.3	28.7	29.4	29.1
32.9	31.8	31.4	28.8	31.1	33.9	35.5	30.7	30.3	30.9	30.8	33.1	31.1	31.2	30.3
33.6	31	30.2	28.1	30.1	34.2	34.7	30	30.3	29.5	30.2	31.9	30.7	30.8	30.5
31.3	29.4	28.2	26.1	27.9	29.9	34	27.8	26	27.6	27.6	30.2	27.4	28.4	28.8
31.9	30.8	29.8	27.2	29.6	32.2	35.7	29.1	26.5	28.5	29.7	32	28.7	30.2	30.5
31.5	30.3	31.2	28.4	30.1	31.5	34.5	29.6	28.7	30	30.5	31.5	29.6	29.9	30.6
32.6	32	31.1	28.3	30.8	33.3	36.2	30.2	29.7	30.6	28.7	32.4	30.8	31.5	31
33.2	32.1	31.2	28.9	30.9	33.1	34.3	29.6	28.9	30.1	30.5	31.3	30.7	31.2	31.3
33.8	31.5	30.7	28.1	30.3	33.6	36.2	29.4	29.5	29.7	30.3	31.9	30.6	30.7	30
32.7	30.9	30.2	27.8	30.3	32.3	35.2	29.5	28.8	29.8	29.7	31.5	30.3	30.4	30.4
31.7	30.3	30.9	27.3	29.4	30.9	34.8	28.5	27	29.2	28	30.8	28.4	28.7	29.2
28.7	27	27.8	25.2	29.4	28.9	29.2	29.1	27.5	27.3	29.1	26.7	29.4	29.3	29.2
29	26.5	27.6	24.5	29.1	29	28.1	28	27.8	27.1	28.4	26.6	28.2	29.4	28.1
25	22.2	24.8	21.4	25.4	25	25	25.3	24.1	23.8	24.5	23.4	26	25.8	25.7
27.9	24.6	26.6	24.4	28	28	27.5	27.7	26.6	26	27	25.6	28.5	28.2	28.5
29.6	26.8	28.4	25.6	30.2	29.7	29.1	29.2	27.8	28	29.3	27.3	29.9	30	30.9
29.2	28.5	30.2	27.1	31.2	29.3	31.1	29.6	29	28.6	30.7	29	30.9	31.4	32
31	29.4	31.1	28.9	32.4	32.3	31.8	31.1	30.4	30.4	31.9	30.3	31.6	32.2	32.8
30	28.4	29.9	27.5	31	31	31.1	29.5	28.8	29.4	30.7	28.9	30.5	30.7	31.1
29.9	29.2	30.1	27.6	31	31.3	31.1	29.8	28.7	28.8	30.3	29.1	29.9	30.9	31.4
30.8	28.9	31.5	28.8	32.8	32.4	32	31.4	29.8	30.8	31.7	29.9	30.7	31.6	31.7
29	27.5	29.1	26.4	30.4	30.2	29.9	29.4	27.1	27.5	29.5	27.4	28.6	30.4	30.8
30.2	28.8	30.9	27.7	32.1	32.1	31.8	30.4	29.5	29.8	31.3	29.2	30.8	31.2	32
31.7	29.7	31.4	28.8	32.6	32.8	32.2	31.4	30.1	30.8	31.9	30	31.1	32.1	32.2
30.9	28.8	30.4	27.8	31.8	32.2	31.1	30.3	28.7	29.2	30.7	29.6	29.8	31.3	31.3
27.4	24.9	27	25.4	28.7	28.5	28.4	27	25.1	26.8	27.5	26.9	26.5	27.9	28.1
28.8	27.7	29.1	25.9	29.9	29	29.6	28.7	28.2	28.5	29.6	27.6	29.2	29.5	30
31	29.6	31.1	28.6	32.1	32.4	32	30.9	29.6	29.6	30.9	30.1	30.4	31.8	31.9
34.5	35.5	33.3	35.6	34	35.6	34.9	33.4	34.2	32.9	36.8	35.4	34.2	35.9	35.6
35.7	33.6	32.8	35.4	32.9	37.9	36.1	33.4	34.9	36.7	39.8	36.5	36.1	37.7	39
33	34.2	32.9	34.4	33.2	34.5	32.9	33.7	34.8	31.8	34.1	34	34.8	34.4	33.7
33.2	33.9	31.5	34.8	33.2	35.3	33.7	32.6	33.4	32.7	35.5	34.2	34.4	36.6	35.9
34.8	35.1	31.4	35.1	33.8	38	36.3	33.9	33.9	35.2	38.5	36.3	35.9	36.6	37
22.6	23.2	23.2	24.3	23.6	25.4	24	22.8	23.8	22	25.3	23.9	24.8	24.2	24.8
33.3	32.3	31.1	34.7	32.4	34.2	33.3	31.7	33.7	34.3	36.6	33.4	33.5	35.2	36
34.6	34.6	31.9	36.2	33.9	36	34.4	33.5	35.3	33.8	37.4	34.7	34.8	36.5	37.5
26.2	24.8	22.5	25.9	26	26.9	25	24.3	26.5	26.1	28.4	25.8	25.4	27.2	28.1
34.9	34.3	32.2	36.3	34.5	37.3	35.1	34.4	35.7	34.6	38	35.3	35.4	37.9	38
31.5	31.6	28.4	32.8	30.8	33.6	31.4	30.4	32.1	31	34.5	32.1	32	33.5	34
29.7	29.5	27.8	29.9	27.9	29.2	28.5	28	28.3	27	30.7	29	29.7	30.4	30.6
36.4	33.1	32.6	34.6	32.9	37.5	35	33	33.5	34.3	37.8	35.5	35.3	36	36.1
31.7	31.5	28.9	32.3	30.7	32.7	30.8	30.5	32.1	30.6	33.6	31.8	32.3	34	34.3
35.5	35	32.4	36.7	33.8	36.8	35.1	34.1	35.5	34.2	37.6	34.5	35.5	36.9	37.3
35.2	33.6	31.1	36.1	33.1	37.2	35.9	33.8	34.2	35.1	38.6	36.5	35.4	36.5	37.7

33.2	32.4	32.8	33.2	32.6	35.9	33.7	31	32.2	33.4	36.7	34.3	33.9	35	35.6
29.7	30.8	30.3	30.5	30.4	31.1	30	30.6	31	28.8	30.9	30.6	32.5	31.1	30.9
29.4	30.9	28.2	31.4	30	32.1	29.3	28.4	30.1	28.6	32	29.7	29.6	31.2	31.2
32	30.8	30.4	32.1	33.6	34.3	33.2	34.8	31.9	30.6	33.5	31.6	32.9	33.2	32.7
30.1	29	27.8	29.9	28.8	31.3	29.6	27.5	28.3	28	32.2	29.8	29.9	30.5	30.9
34.9	34.2	32.3	36.3	34.3	36.1	34.7	33.8	35.7	34.8	37.9	35	35.8	36.8	37.6
33.2	32.4	31.5	33.8	32.3	35.8	33.1	32	32.6	32.4	35.8	33.8	33.6	35	34.8
32.5	31.8	31.3	32.1	32.2	34.7	33.5	30.6	30.8	31.1	34.8	32.6	31.5	33.5	32.1
36.5	33.7	34.2	35.7	32.7	38.2	35.3	33.2	35.8	36.8	39.7	36.2	36	36.9	38.8
26.7	27.8	26.3	24.9	27	27.2	26.9	26.7	26.4	24.3	27.6	26.6	27.9	27.3	27.5
29.4	28.5	28.5	29.4	28.2	30.4	29.5	26.9	28.7	26.8	31	28.3	29.8	28.4	29.2
34.9	34.3	32.6	36.2	34.1	36.8	34.6	33.7	34.8	35.1	37.8	35	35.4	36.8	37.5
30.3	31	30	31.4	30.5	32.4	31	29.7	30.3	28.9	32.8	31	31.1	32.6	32.4
35.9	37.3	36.1	34.3	35.1	33.6	34.5	36.4	36.9	32.6	34.8	35.7	36.6	32.8	35.2
34.8	35.9	33.9	33	33.4	33.1	34.2	35.3	35.2	30.9	33.7	35.8	36.2	33.2	34.4
34.5	35.5	34.3	34.3	34.5	33.7	34.1	35.7	35	31.8	34.2	35.5	36	33	33.9
35.2	35.7	34.6	34.2	34.3	33.1	34.3	35.4	35.3	31.5	33.9	35.9	36.3	33.5	34.8
36.9	37.5	35.5	35.5	35.1	34.5	36	37.1	36.7	32.7	35	36.8	37.4	34.1	35.8
36.8	37.2	36.7	36.6	36.1	35.3	35.1	36	35.9	32.1	36.8	36.8	37.5	35.7	37
31.4	32.4	32	31.7	32.6	30.1	31.7	31.9	30.9	28.6	31.9	32.2	32.2	31.7	32.3
39.2	39.5	37.5	37	36.9	37	38.6	39.3	38.4	34.4	37	38.5	38.8	35.4	37.6
32.3	33	31.4	30.2	30.4	30.1	31.8	32.5	32.5	28	31	32.7	33.2	30.2	31.1
35	35.3	33.6	33.3	33.2	32.6	34.1	35.3	34.8	30.5	33.1	34.8	35.3	32.1	33.5
30.3	31.1	30.2	29.4	30.1	29	30.3	29.9	29.9	25.6	28.4	29.5	31.4	28.2	29.9
37.7	38.1	37.7	38.1	37.5	36.2	36	36.2	36.2	34.4	37.5	37.1	37.5	37.1	38
39.7	40.3	38.4	39.9	39.8	37.9	37.1	37.5	38.5	36.5	39.1	39.2	38.6	39	40.2
28.7	27	28.1	25.5	30.6	29.6	29.5	27.4	26.3	27.1	28.1	26.2	28.4	28.6	29.1
23.9	20.7	22	19.7	23.6	23.5	23.8	24.8	22.6	22.7	23.3	21.3	25.5	24.8	25.7
29.9	28.2	29.7	26.6	31	30.7	30.1	30.6	29	29.3	30.7	28.8	30.5	30.8	31.1
27.2	26.1	27.7	25.3	29	28.5	28.1	27.6	25.6	27.2	27.7	26.4	27.1	28	28.1
29.6	28	29.6	25.9	29.9	29.5	29.8	28.9	27.6	27.7	29.5	27.6	30	29.9	30
28.6	27.2	28.1	25.4	30.2	31.4	30.5	28.9	27	27.1	30.9	27.4	29.5	29.4	31.5
25.4	23.4	24.2	22.6	26.4	27.9	27.1	25.9	24.4	24.6	26	24.8	27	26.7	27.8
27.2	25.2	26.6	24.2	28	28.9	28.7	27.2	25.8	25.6	28.7	26	28	27.7	29
29.5	28	30.1	26.9	32.3	31.6	30.7	30.2	28.6	30	30.1	28.7	29.4	30.6	30.8
29.6	28.1	29.2	26.5	30.9	29.9	30	29.4	28.3	28.8	30.2	28.5	29.9	30.2	30.5
29.7	27.5	28.8	25.7	30.2	30.3	30.2	29.9	27.8	28.7	29.4	27.4	30.8	30.4	30.5
28	26.3	27.2	24.9	28.9	29.7	29.4	28	26.4	26.6	29.1	26.4	28.9	28.4	29.8
28.3	27.7	27.6	26.1	29.6	29.5	29.3	29.9	28.2	28.9	29.8	28.3	29	30.3	29.8
27.9	27.8	29	24.4	30.3	30.8	29.7	28.5	26.4	27	30.1	27.3	28.4	28.8	30.4
27.5	25.8	27.3	24.3	28.3	29.5	27.8	26.3	24.5	25.8	28.1	25.4	27.2	27.9	29.3
30.7	27.7	30	27.8	31.4	28.6	31.5	31	29.4	30	30.8	28.3	31.5	30.9	31.6
26.5	25	26.2	23.5	28.1	28.6	28	26.8	25.1	25.1	28.1	25	26.8	27.6	28.7
26.5	24.7	25.9	23.7	27	27.2	25.1	26.7	24.8	25.4	26.2	24.6	26.2	25.7	28.1
28.7	25.4	27.6	25.2	28.7	28.9	28.9	27.9	26.3	27.2	27.6	26	30	29.5	29.6
30.1	28.6	30.2	27.4	31.1	29.9	30.4	29.7	28.9	29.4	30.7	28.8	30.3	30.7	31
29.2	27.4	28	25.9	29.9	30.6	31.1	28.9	27.6	28.2	29.5	27.1	30.1	29.8	31.1

27.3	24.9	26	23.3	27.7	28.5	28.6	26.2	24.2	25.5	27.7	25.7	28.8	27.6	28.6
27.1	24.8	26.2	23.7	27.9	28.1	29	27.6	25.3	26	26.9	25.8	27.8	27.6	29.6
28.9	27.6	28.5	25.5	30.1	29.2	30.1	29.5	27.5	28.3	29.5	27.5	29.8	30.4	30.4
28.5	27.4	28.8	26.1	30.1	29.7	29.8	29.6	28.7	29.4	30.8	28.5	29.6	30.4	30.4
26.4	24.6	25.3	23.1	26.6	27.2	27.9	26.7	25	25.4	27	24.5	28.4	26.6	28.3
27.4	24.9	25.7	23.6	27.7	28.4	28.2	26.7	24.9	26.5	27.1	24.9	29	28.4	29.7
27.6	25.5	26.2	24.2	28	28.7	29	27.4	25.7	26.4	27.9	25.8	28.9	28	29.3
27.3	25.6	26.3	23.9	27.6	28.8	28.6	26.5	25.8	25.4	28.5	25.4	27.9	27.4	28.5
25.8	23.7	24.6	23.2	26.9	26.8	27.2	26	23.9	25	25.8	24	27.8	26.5	27.7
26.8	25.2	25.9	22.9	27.1	28.2	28.2	26.2	24.1	25.4	28	25.2	28.2	27.5	28.1
27.7	26.5	26.9	23.9	29.2	29.2	29.4	28	26.3	26.5	28.9	26.7	28.4	28.5	29.8
28.7	26.9	27.5	24.7	28.4	28.9	29.9	28.6	26.6	27.3	28.3	26.5	29.4	28.5	29.9
28.2	25.3	26.2	24.3	28	28.1	29.2	27.3	24.8	26.8	27.8	25.6	29.6	28.7	30.1
27.4	24.7	26.2	23.9	27	28	28.6	27.2	25.1	25.7	27.5	25.2	28.8	27.5	29.3
29.9	28.3	30.2	26.4	31.3	30.7	30.5	30.1	28.8	29.4	30.5	28.6	30	30.7	31
26.4	25.1	26.3	22	27.5	29.1	28.2	26.1	24.1	25.7	26.9	25	27.6	26.5	28
29	27.2	28.2	25.9	29.9	31.5	30.5	28.3	26.7	26.8	30	27.3	29.6	29.2	31.1
29.2	27.2	28	25.9	29.6	29.7	29.9	29.5	27.8	28.3	29.2	27	30.2	29.4	30.6
27	25.5	26.2	23.3	27.4	27.9	28.6	26	24.1	25.8	27.8	25.4	28.1	27.4	29
28.2	26.1	27.1	23.9	28.5	29.5	29.5	27.2	25.5	26.4	28.6	26.2	29.1	28.1	29.7
24.6	22.5	23.4	21.5	24.9	25.9	25	24.8	23.3	23.7	24.7	22.4	26.4	25.1	26.2
26.6	24.8	25.6	22.9	28.3	29.1	27.4	27.2	25	26.8	29.5	26.8	28.5	28	29.1
27.6	26.1	26.7	23.7	28	29	29.5	27	24.9	26.7	28.3	26.2	28.8	28	29.9
27.7	25.7	26.8	24.1	28.4	28.2	28.7	27.7	25.9	26.2	27.5	25.6	28.7	27.9	28.7
26.6	25	25.6	23.6	27.2	28	29.1	27.5	25	25.1	27.5	24.5	28.4	27.3	29.1
25.8	23.8	25.8	22.5	26.6	26.5	27.7	26.4	24.4	24.5	25.6	24.2	27.3	26.7	27.3
28	25.5	25.6	24	27.7	28.3	29.3	27.6	25.5	26.7	28.2	25.3	30.3	26.2	29.5
27.7	25.6	26.8	24.7	28.5	29.2	29.2	27.5	25.8	26.5	29	25.8	28.4	27.6	29.3
24.7	22.7	23.7	21.4	25.3	25.7	25.9	24.8	23.1	24	24.7	22.4	25.7	25.5	25.8
28.5	27.1	27.1	26	30.5	29.8	30.4	29.5	27.8	27.8	29.3	26.8	29.8	29.6	30.3
26.6	24.2	25.5	24.1	26.8	27.6	26.9	26.2	24.8	25.6	26.5	24.5	28.5	27.4	28.6
28.6	26.7	27.7	24.2	29	30.2	30.1	27.4	25.5	26.7	29.5	26.6	29.8	28.5	29.9
27.1	24.1	25	22.7	26.6	27.2	27.4	26	24.5	25.4	25.8	23.9	27.6	27.6	27.5
27	25.4	26.4	23.5	27.7	29	29	26.5	24.8	26.2	28.5	25.9	28.7	27.7	29.1
27.7	25.7	26.6	24.2	28.1	29.2	28.7	27	24.8	25.7	28.5	25.8	27.9	27.7	29.2
28.3	25.8	26.7	24	27.5	28.2	27.9	26.9	24.6	25.4	27	25.9	27.8	27.5	27.1
28.1	27.5	27.6	25.5	29.1	30.2	30.9	27	25.5	27.2	29.1	27.6	28.7	29.1	30.7
28.2	27.5	26.9	24.3	28.6	30	30.1	27.1	25.7	27.2	29.6	27.6	28.2	28.3	29.7
27.6	26.2	26.9	24.7	28.9	29.3	29.9	26.6	25.8	26.8	28.4	27.1	28.2	28.2	30
27	26.8	26.8	24.1	28.4	29.3	29.5	25.6	24.7	26.4	28.5	27	27.6	28.1	28.8
27.6	26.4	26.8	24.7	28.6	29.9	30.1	26.8	25.5	26.5	28.8	26.9	27.6	28.6	29.5
28.4	27.5	28	26.1	29.7	31	31.8	27.9	26.4	27.8	30.1	28.3	29.4	30.1	31
30.7	29.8	30.3	27.5	31.3	31.8	32.7	29.8	29	29.3	30.1	30.1	30.6	31.5	33
28.3	27.5	28	25.9	29.6	30.8	31	27.3	26.3	27.4	30	27.8	29	27.8	30.2
29	27.6	28.6	25.5	29.8	31.4	31.8	27.8	26.8	27.9	29.4	28.9	29.4	30.6	31.7
29	27.8	28.4	26	29.6	31.4	31.3	27.8	26.5	27.8	30.7	28.2	29.4	29.8	30.6
29.3	28.2	28.8	26.4	31.1	32.5	34	27.5	26.7	27.2	30	28.2	29.4	30.4	31.6

30.5	28.9	30.2	27.1	31.4	32.2	33.3	29.2	28.5	28.7	31.1	30.2	30.5	31.6	32.9
28.5	27.1	28.5	26.1	30.3	31.5	31.4	27.6	26.6	27.2	30.1	27.8	29.6	30.3	31
29	28.6	28.7	25.9	30.3	31.5	32.1	27.7	26.5	27.5	29.9	28.2	28.8	30.2	31.2
29	27.8	27.9	25.7	29.4	30.6	30.6	27.3	25.9	27.2	29.9	27.9	29.1	29.7	30.4
28.8	27.5	28.2	26.1	29.9	31	32.1	27.2	26.3	27.6	29.5	28	28.9	29.5	30.9
29.6	27.5	28.6	25.7	30.6	31.9	31.9	27.9	27.6	27.7	29.2	27.8	28.6	30	31
29	27.7	28.6	26	30.2	31.9	32.2	27.6	26.5	27.7	29.7	28.4	29	30.5	31
29.9	28.6	29.4	26.8	30.8	32.3	32.2	28.3	27.2	28.4	30.7	28.9	29.6	30.9	31.5
29.4	27.2	28.9	26.2	30.4	31.8	32.4	28.1	27.1	27.9	29.5	28.8	29.7	31.2	31.7
27.9	27.7	28.2	26	29.9	30	31.4	28	27	28.1	29.1	28.5	29.4	30	31.4
28.7	27.8	28.3	25.7	30.1	32.2	33.1	27.2	26.6	26.9	29.9	28	29.4	30.2	31.2
29.8	28.6	29.5	27	30.9	32.4	32.9	28.4	27.3	28.1	30.9	28.8	29.2	30.6	31.1
29.1	28.1	28.8	26.3	30.5	31.9	32	27.4	26	27	29.4	27.8	28.3	30.1	31.2
28.8	28	28.9	25.9	30	31.3	31.9	27.6	26.4	27.2	29.7	27.9	28.6	30.1	30.8
29.2	27.7	28.5	26	30.2	32.8	32.8	27.6	26.2	27.6	29.4	28.6	28.8	30	30.2
38	33.5	35.8	35.8	34.2	40.6	39	34.9	33.1	36.2	35.9	36.7	37.5	36	35.8
33	33	33.2	33.2	32.8	35.8	35.2	32.2	32.4	32.6	35.9	33.2	33.3	34.5	33.8
35.6	34.4	33.6	35.4	34.1	39.2	36.2	33.8	33	34.2	37	35.2	33.7	34.9	35.2
32.6	32.6	32.4	32.9	32.1	36.5	34.6	32	32.3	32.8	35.4	33.6	33.5	34.4	33.8
36.7	32	35.1	35.2	33.3	40.5	38.1	33.4	32.7	34.7	35.4	36.1	36.2	34.7	35.1
38.7	33.1	36.2	36.3	32.9	41.2	38.5	34.9	33.9	35.6	37	36.9	38.1	36.3	37.6
36.1	30.9	34.4	33	33.6	38.4	36.3	32.7	31.4	34.2	35	33	35.2	32.5	34.1
36.7	32.3	34.7	34.8	32.9	39.9	37.7	33.6	32.5	34.3	35.3	35.5	36.7	34.8	35.8
37.3	33.3	35.9	34.8	35.1	40.8	39.1	34.7	32.1	35.5	35.5	35.3	36.1	34.3	35.2
36.5	31.4	34.2	35	33.5	39.6	38.1	32.8	31.5	33.8	34.9	35.3	35	34.3	34.2
37.9	32.8	36.7	35.1	34.2	40.9	38.1	33.4	33	34.6	35.5	35.4	35.9	34.3	34.3
36.7	31.4	34.7	35.6	33.9	40.2	37.8	34.2	33.3	35	36	35.5	38	34.6	35.3
37.2	32.2	35.2	34.7	34.6	40.3	37.4	33	32.5	33.9	35	34.8	36	33.9	34.2
36.8	31.8	34.5	35.5	33.9	40.2	38.6	33.4	32	34.3	35.3	35.9	35.5	34.9	34.8
36.9	30.9	33.1	34.7	32.6	40.7	37.8	32	32.1	33.1	34.5	34.9	34.6	33.9	33.7
38	31.8	35.9	34.7	33.5	40.1	37.1	32.5	32.5	33.7	34.8	34.8	36.6	33.7	34.1
36.8	32.1	36.2	35.8	33.8	39.6	37.1	32.9	32.5	34.5	35.3	34.6	37	33.7	34.3
36.4	31.6	35	33.6	34.1	39.3	37.5	33.1	31	34.5	34.5	33.8	34.8	32.7	33.8
35.4	31.6	34.1	33.5	33.1	39.5	37.2	32.3	31.3	32.9	33.9	34.1	34.4	33.5	33
36.6	31.7	34.7	35.5	33.6	41.3	38.8	33.3	32.5	34.4	35.5	35.9	35.6	35	35
35.5	33.4	34.1	35.4	35.2	39.5	37.8	33.2	32.8	34.7	36.4	35.4	33.9	34.7	35.5
36.8	32.8	34.5	34.9	34.2	40.5	39.1	33.6	32	34.4	35.3	35.6	35.3	34.7	34.6
35.8	30.2	32.8	33.3	32.2	38.9	36.6	32	31.2	32.5	33.7	33.8	33.8	32.8	32.7
36.3	32	35	33.9	33.9	40.1	37.9	33	31.7	33.7	34.7	34.5	35	33.6	33.9
36	31.4	33.9	35	33.1	40.3	38	33.2	32	34.4	35.5	35.8	35.9	34.2	35.4
36.2	32.8	33.8	34.8	33.9	39	36.9	33.3	32.7	33.7	35.6	35.8	34.5	34.3	35.3
38.5	31.6	34.2	36.7	34.2	40.2	38.5	33.6	31.9	34.5	35.5	35.7	34.9	34.6	35
37.7	32.1	35.9	35.4	33.6	40.8	37.7	33.7	32.9	34.4	35.7	35.4	37.1	34.4	34.9
35.8	32	33.7	34	33.6	39.8	38.6	32.8	31.3	33.5	34.3	34.6	34.5	33.7	33.8
37.4	32.8	34.5	35.9	34.2	40.6	39.3	34.1	32.4	35	35.8	36.3	35.8	35.4	35.5
36.5	32.5	35.2	35.3	34.7	39.3	38	32.8	31.2	33.9	34.6	35.4	34.9	34.2	33.8
35.4	31.1	33.5	32.6	33	38.3	37.6	32.7	30.6	33.6	34	33.2	34	32.2	33.4

37	31.9	35	36.1	34.5	39.6	36.5	33.2	32.7	35	35.8	34.4	36.7	33.5	34.7
37.4	32.4	34.6	35.5	33.7	40.2	38.6	33.9	32.2	34.6	35.8	36.2	35.9	34.9	35.9
37.2	32.2	35	35	32.7	40.2	37.6	33.9	32.9	34.7	35.8	35.4	36.9	34.9	35.6
36.8	32.8	34.5	35.2	34.3	40.3	38.9	33.5	32	34.3	35.2	35.6	34.7	34.6	34.2
36.3	31.9	34.9	36	33.7	40.6	38.2	33.3	32.8	34.3	35.3	36	35.5	34.9	34.5
35.7	32.4	33.4	32.8	33.3	38.6	38.6	32.9	30.5	33	33.8	34.3	33.7	32.6	33
34.7	31.4	32.7	32.2	32.7	38	37.8	32.7	30.4	32.8	33.2	33.3	34.4	32.3	33.2
34.5	31.4	32.7	32.3	32.6	38.5	37.9	32.4	30.2	32.8	34.3	33.1	34.5	32.5	33.2
34.5	32.1	32.6	31	32.7	38	38.3	32.7	30.4	32.6	33.3	33.3	34.2	32.1	33.7
34.7	31.3	32.5	31.9	32.8	38.1	38.1	32	31.1	32.5	33.2	34.1	33.3	32	32.5
35.6	30.9	34	34.8	32.2	38.4	37	32.7	30.8	32.6	34.4	34.5	34.4	33	35.2
35.7	31.3	32.5	31.5	32.3	37.3	37.6	31.9	30.6	32.3	33.2	33.8	32.8	31.9	32.3
30.3	30.4	28.6	30.1	28.4	26.1	28.5	31.2	30.8	28.8	26.9	31	31.5	26.8	29.7
34.5	32.5	33.5	35.4	33.3	29.5	32.2	35.5	35.5	35.9	32.7	34.5	34.3	32.5	33.1
31.3	29.5	30.5	31.3	29.8	27	28.4	31.6	31.5	31.4	27.4	30.6	32.1	27.5	28.5
32.8	33.8	30.6	30.9	28.7	28.3	31.6	32.7	33	30	28.6	33.1	32.2	29.3	30.8
29.4	30.5	27.7	28	26.6	25.6	28.4	30.2	29.7	26	26.7	29.8	29.5	26	27.1
30.7	31.7	29.7	29.7	29.3	27.5	29.1	31.1	31.3	27.6	28.4	31	30.8	28	29.3
32.2	30.8	31.6	33.5	31.6	27.8	28.8	33.8	32.4	34.2	30.1	31.8	33.5	30.3	31.3
32.2	32.2	31	32.9	30.4	27.8	30.5	33.7	33.2	32.2	29.3	33	33.8	28.6	30.6
28.2	27.8	27.3	29	26.1	23.9	25.8	28.3	28.5	29.6	25.6	28.6	30.2	26	27.5
33.2	33.8	30.9	30.3	29.4	29.4	31.8	33	32.5	29.1	29.9	32.7	32	29.6	29.5
30.5	30.1	29.1	31.2	28.6	25.9	28.6	31.8	31.4	30.8	27.4	30.5	31.6	25.6	27.1
31.9	30.5	31	32.8	31.3	28.2	29.6	33.6	33.5	31.7	29.9	32.4	33.1	29.9	31.3
36.2	36	33.8	32.3	32.1	30.3	34.6	36.5	35.3	32.4	32.1	35.8	34.7	31.9	32.4
33.2	33.9	30.2	31	29.5	28.8	31.3	33	33.5	30.7	28.7	33.9	32	29.4	30.3
21.9	22.4	20.9	22.5	20.7	16.8	19.6	24.1	23.5	20.9	19.2	22.2	23.2	19.1	20.8
30.2	29	29.4	30.7	29	25.6	26.5	31.4	30.3	31.9	27.8	29.9	31.2	27.9	29.4
32.4	33.2	29.3	30.5	28.4	27.5	30.6	31.2	32.3	29.7	28.1	32.1	31.6	28.6	30.6
27.2	25.7	26.3	28.2	25.8	23.2	24.3	27.6	27.9	28.5	24.8	27	28.5	25.2	26.2
29.8	28.4	28.7	29.9	27.9	25.4	26.6	30.5	30.3	31	27.1	29.2	31	27.2	28.3
31.3	32.4	31.1	30.3	30.5	28.8	30.1	32.2	31.8	28.1	29	31.5	31.5	28.7	30.3
30.9	31.2	29.6	30.2	28.9	26.3	29.1	31.7	31.9	30.8	27.4	30.7	31	27.4	28.8
34	32.5	32.4	33.6	32	28.1	31.4	34	34.2	30.6	31.1	33.5	33.3	30.3	31.5
29.5	28.5	28.4	30.8	27.7	26.3	26.8	30.1	30.7	31.4	27.3	29.9	30.9	27.3	28.9
33.7	32.3	31.4	32.7	31.1	29	30.9	33.2	33.5	32.9	29.5	32.8	32.9	29.3	31.2
31.6	30	30.3	31.8	30.3	25.8	27.9	31.6	32.5	30.2	28.3	31.5	31.6	28.9	29.8
34.8	34.1	32.7	34	32	29.7	32.1	34.7	35.1	34.5	31.3	34.4	34.1	31.3	32.6
26.7	26.3	25.8	28.2	25.2	21.8	23.6	27.6	27	28.1	24.5	26.5	28.3	24.1	24.2
29.5	28.8	28	29.7	27.2	24.4	25.2	28.9	28.7	29.7	25.9	27.9	29.5	26.1	27
31.7	31.6	29.7	31.2	29.6	27.2	29.6	32	31.8	29.1	28.2	32	32	28.4	30.3
32.1	31.4	31.8	33.6	31.7	28.1	29.3	33	33	33.9	29.5	31.6	33.4	30.2	31.1
34.1	34.7	32.9	34.3	31.7	29.7	32.2	34.8	35.3	33.4	30.4	34	34.2	31.6	32.7
24.9	24.2	23.6	26.7	22.9	21	21.9	24.4	25	27.3	21.9	25	28	22.4	23.7
32.1	31.9	30.2	32.1	29.1	27.4	30.4	32.2	32.4	30.6	28.3	32.7	32.5	28.5	30.1
28.8	27.1	26.8	29.7	26.9	24.3	25.8	28.5	29	29.8	25.5	28.1	29.9	25.8	26.7
32.5	32	30.6	31.6	29.7	27.3	29.7	32.7	32.7	31.4	28.7	31.7	31.8	27.9	29.5

18.5	19.1	17.7	17.5	17.2	17.7	17.2	18.4	18.8	20.2	18.3	18.6	18.7	19.7	18.6
19.8	20.4	18.8	20.3	18.7	18.9	18.4	18.8	20.4	22.4	20	20.2	20.2	20.4	19.5
19.3	19.7	18.4	19.6	18	18.4	18.2	18.3	19.9	20.9	19	18.8	19.3	19.5	19.3
19.3	20.2	18.1	18.2	17.3	18.3	18.1	18.4	18.8	20.7	19.5	18.8	18.9	20.4	19.1
16.3	18.4	16.5	16.4	15.5	16.9	16.4	16.8	17.6	19.3	17.8	17.3	17.1	18.9	17.5
25.8	25.1	25.5	22.9	27.7	30	28	26.3	25	25.5	28.8	25.8	26.5	27	28.3
26.2	23.8	25.4	22.6	27.3	26.9	27.1	26.4	24.6	24.5	26.2	24.9	27.1	26.7	27.3
29.9	29.6	29.1	26.9	31.5	32	31.5	29.8	28.3	28.9	31.8	29.7	29.3	30.3	32.2
30	29.7	30.7	26.7	31.6	33.9	31.5	29.5	28.4	29.9	32.1	30.7	30.9	31.2	32.6
30	28.4	29.8	26.9	31.1	31.6	31	29.8	28.5	29.1	30.6	28.8	30.1	30.4	31.5
31.4	29.7	30.5	28.1	31.9	32.5	31.9	30.7	29.8	30.1	31.6	30.1	30.4	31.4	32.5
30.3	29.4	29.9	27.1	32	31.7	31.7	29.8	29	28.7	32	29.5	29.6	31	31.9
31	29.3	30.3	28.3	31.6	33	31.8	30.3	29.8	29.6	31	30	30.1	31.6	32.1
27.6	25.2	26.7	23.8	28.3	28.9	28.5	27.5	25.7	25.8	28.1	25.9	28	27.6	29
30.2	30.4	30.1	28.1	32.4	32.4	31.6	30.1	29	29	31.6	30.2	30.4	31.5	32.9
29.7	28.6	29.1	26.9	31.5	31.8	31.2	29.7	27.9	28.3	30.8	28.6	28.9	30.1	31.6
31	31.1	30.9	28	32.7	33.8	32.4	30.5	29	28.9	31.3	30	29.5	31.2	32.2
29.6	27.9	28.7	26.5	30.1	30.4	30.3	28.8	27.5	27.8	30.6	28.2	29.2	29.7	30.9
33.5	29.9	30.6	28.6	31.9	32.8	32	30.5	29.5	29.5	31.6	30.8	31	31.6	32.9
27.6	26.7	27.2	24.3	29.5	31.1	29.7	27.5	25.7	27.1	28.8	26.8	28.4	28.4	29.8
28.4	27.1	28.2	24.6	29.5	30.7	29.9	28.2	26.2	27.6	28.8	28	29.2	28.9	30.5
32.3	30.7	31.8	29	33.3	33.8	33.5	32.1	30.5	31	32.8	30.8	31.8	32.6	33.5
28.5	27.3	28.5	25.4	29.5	31.1	30.6	28	26.5	27.8	29.4	28.4	29.4	29.6	30.7
28.1	26.8	27.8	25	29.5	30.7	29.6	27.7	26.3	26.7	28.9	27.2	27.4	28.8	29.9
27.2	25.8	27.3	23.4	27.9	29.1	28.6	26.3	24.7	26.2	27.8	26	27.8	28	28.8
30.5	29.5	30.5	27.9	32.3	33.3	32.7	30.5	29.2	30.3	31.9	30.4	30.8	31.3	33.3
28.8	27.8	28.7	26.5	30.4	30.5	31.5	28.5	27.8	28.5	29.7	29.1	29.4	30.3	32.2
27.5	26.5	27.4	24.2	28.7	30.1	29	27.4	25.3	26.9	27.9	26.9	28.6	29.5	29.7
29.5	27.9	29.1	26.6	31.2	30.7	30.8	29.3	27.8	28.3	30.1	28	29.9	30.1	30.4
27.8	27.1	28.2	25.9	29.7	29.7	28.8	29.4	26.4	27.7	28.6	26.7	28.3	29.5	28.8
26.6	25.5	26.5	24	28.3	27.7	27.1	27.4	25.4	26.2	27.2	25.4	26.5	27.9	27.5
28.6	27.6	29	25.3	30.1	29.4	29	29.6	27.7	28	29.2	26.8	28.9	29.5	29.7
34.6	33.3	33.3	33.8	35.3	35.6	35.8	30.6	33.2	35.5	37.1	34.1	33.8	34.8	34.1
33.1	32.4	32	32.6	34.6	34.5	33.7	30.9	32.2	34.4	35.2	33.1	32.5	34.3	34.5
33.7	31.9	34.2	32.1	34.4	34.9	35.6	30.2	31	35	36.2	34	32	34.5	33
30	29.4	32	30.6	32.7	32.7	33	29.3	29.8	31.6	34.4	31.4	30.4	31.9	32.9
31.6	30.2	32.1	31	33.4	33.7	34.1	29.8	31.5	33.5	34.3	32.4	31.4	32.2	32.7
32.3	30.2	32.8	31.1	33.5	33.4	33.6	29.6	30.5	32.5	35	32.7	31.1	32.4	33.1
33.4	32.2	33	32.3	34.1	34.8	35.2	31.4	32.5	34.2	34	32.9	33.5	34.3	33.5
33.9	32.2	33.4	33	34.2	35.5	34.9	31.1	33	35.1	35.5	33.5	32.7	34.2	34.1
32.9	31.1	32.8	31.4	34.1	34	34.2	30.2	30.9	32.7	35.1	32.6	31.5	33.3	33.4
33.5	30.7	32.4	32	34.7	33.9	34.4	30	31.3	33.4	35.9	32.7	31.2	33.1	33.3
30.6	30.2	30.2	30	31.1	31.1	31.4	29.3	30.7	31.7	32.4	30.8	30.6	30.9	31.4
33.4	32.3	34	32.8	34	34	33.7	31.3	32.2	33.7	35.1	33.3	32.4	33.7	33.7
32.8	31	33.4	31.3	34	34.6	33.3	30	31.2	33.5	34.6	32.9	31.6	32.9	33.2
31.6	30.3	30.7	31.3	31.9	33.6	33.9	29.9	30.6	33.1	33.2	32.2	32.3	33.3	32.3
31.9	30.5	31.6	31.3	32.6	34	34.1	29.8	30.9	33.4	33.7	32.3	31.6	32.7	32.2

31.8	30.8	32.2	31.2	33.9	33.9	34.4	29.6	31.1	33.6	34.8	32.2	31.5	32.9	33
33.3	31.5	33.5	33	35.3	35	34.3	30.1	31.7	34.1	35.9	33	32	33.5	33.3
33.5	31.7	32.7	32.2	33.5	34.3	33.9	31.3	32.9	33.9	35.4	32.9	32.4	33.7	33.7
32.8	33.5	33	32.4	33.4	33.8	33.8	30.8	32.2	33.9	34.7	33	32.4	33.5	33.2
32.9	31.5	32	31.9	33.5	34.8	34.8	30.8	31.3	34.2	34.5	33.1	32.1	33.1	33
33.8	31.8	33.3	32.8	34.6	35.3	34.9	30.7	32.9	34.6	36.1	33.4	32.8	34.4	34.1
32.2	31.2	31.4	31	32.2	32.8	33.2	30.1	30.5	32.5	33.5	31.5	31	32	31.9
34	32	33.9	33.3	34.8	35.4	35.6	30.9	32.2	34.8	36	33.9	33	34.4	34.1
32.2	30.7	33.3	31.4	33.8	33.8	33.6	29.6	30.1	33.7	34.5	32.5	30.9	32.5	33.2
34.3	32.6	34.7	33.1	35.5	35.5	35.7	30.7	33	34.6	35.9	33.6	33	34.5	34
33.9	32.8	33.5	33.4	34.6	35.4	35.1	31.3	32.9	34.8	35.9	33.6	33.6	34.3	34.1
32.5	30.5	32.3	31	32.8	34	33	29.7	31.3	32.9	33.9	32	32.2	33	32.9
33.8	33	33.6	32.9	34	34.9	34.9	30.7	31.5	33.7	35.3	33.3	32.6	33.8	33.8
31.5	30.8	32.2	30.6	33.3	33.1	33	29.7	30	33.1	33.6	32.1	31.1	32.5	32.6
32.3	32.4	28.9	26.4	27.6	29.7	32.7	28.8	27.4	27.8	29	31.9	26.6	27.3	29.1
35.2	33.2	31	27.7	30.8	31.4	35.3	30.1	29.7	30.5	31.5	34.3	30.5	29.2	30.6
36.5	35.3	30.7	29.3	31.2	32	35.9	31.4	30.5	30.4	31.8	34.7	31.2	28.7	32
32.4	30.5	29.1	25.3	28.9	31.2	33.4	27.4	26.4	28.1	29.2	31.1	27.3	29.4	29.6
34.2	33.7	30.1	28.6	29.8	32.4	34.7	29.6	29.3	28.9	31.8	33	29.3	28.8	32.1
35.2	32	30.8	27.3	30.3	31.5	35.6	28.5	29.8	30.3	31.4	34	29.5	29.6	29.9
30.5	30.5	29	25.6	28.7	30.7	33.3	28	26.2	29	28.6	30.5	28.6	28.9	30.3
36	33.6	31.7	28.2	31.2	33.2	36	29.7	30.7	31.9	32	35.9	30.9	29.9	31.3
38	35.1	32	29.3	31	33.1	35.4	32.6	29.5	32.1	34.1	36.1	32	29.6	32.3
36.2	34.5	30.8	28.8	30.9	31.3	35.9	30	29.5	30.9	32.3	34.5	31.4	29.4	32.1
33.8	31.1	29.3	26.1	28.6	31.2	34.6	28	28.3	28.2	30.3	31.7	28.6	29.3	30
33	29.2	28.2	24.9	28.4	29.1	31.3	27.3	26.2	28	27.2	29.8	26.8	26.7	28.3
33.7	30.9	30.6	27.3	29.8	31.4	32.5	28.7	27.3	29.9	29.4	31.2	29.6	29	30.9
35	32.5	30	27.5	29.9	30.6	34.4	28.8	28.7	30.1	30.8	33.8	29.6	29	30
33.1	31.4	31.1	27.4	30.1	31.8	35.3	29.2	28	28.9	30.3	30.9	28.4	29.9	29.7
36.9	34.1	31.4	28.6	30.9	32.1	35.3	30.7	29.4	31.3	32.2	34.8	31.2	29.9	31.8
32.8	29.3	29.6	26.1	29.4	31	32.2	28.5	26.9	29.1	29.1	30.2	28.4	28.5	30.1
31.8	30	28.1	24.7	27.8	29.5	31.5	27	25.5	28.1	28.9	28.8	27.1	27.4	28.6
32	29.8	28	24.7	27.4	29.1	31.4	27.1	25.5	28	28.2	29.7	27.4	27.5	28.8
35.6	33.7	31.2	27.3	30.8	30.8	35	30.3	29.2	29.8	31.2	35.2	30	29.9	31.3
34.4	31.6	29.6	26.1	29.1	30.7	32.6	28.5	27.6	29.6	29.3	31.4	28.8	28.5	29.5
32	30.2	29	26.1	29.4	31.4	35.1	29.1	26.4	27.9	28.9	31.2	28	28.8	29.9
34.1	30	29	25.9	28.5	29.6	32.1	27.4	26.2	29.3	29.3	31.2	28.6	28.5	29.9
34.4	31.9	28.9	27.5	28.6	30.4	32	28.1	27.3	29.2	29.1	33	27.2	28.2	28.5
32.6	30.1	28.6	25.3	27.8	30.2	32.8	27.1	25.7	27.5	28.2	29.5	28.1	28.2	29.2
28.7	28.1	28.3	25.7	30.2	29.1	30	26.1	26	27.6	28.7	28.5	28.5	27.7	28.3
33.5	31.5	33.2	30.5	34.3	33.5	35.4	30	30.1	32.5	34	32.9	31.7	33	33.2
33.7	32.6	33.7	30.2	33.8	34.3	34.7	30.9	30.2	33	33.8	33.5	33.3	32.3	33.6
32.5	30.4	31.6	29.1	32.6	32.1	33.2	29	29.1	31	32.4	31.3	30.8	30.8	31.6
33.4	33.4	33	29.7	34.7	34.6	35.5	30.9	31.3	32.4	33.6	34.1	33.2	33	33.7
32	30	31.5	29.1	32.7	32.5	33	28.5	29.2	30.5	31.7	30.7	30.9	31.1	31.9
31.8	29.9	31.2	28.8	32.9	32.2	32.2	29.4	29.7	30.5	33.2	31.8	31.6	31.2	32.7
30.7	29.1	29.7	27.7	31	30.3	31	28.2	28.1	29	31.4	30.2	30.4	29.7	31.6

32.9	31.6	32.3	29.8	33.2	33.3	33.8	29.4	29.7	31	32.9	31.6	32.4	31.8	32.8
31.9	29.4	31.4	29.7	32.8	32.2	32.6	28.7	29.2	30.9	33.2	30.8	30.6	31.3	31.9
31.4	32	31.7	28.8	32.5	33.2	34.3	28.6	29.9	31	32.4	30.8	31.2	31.2	31.5
32	31	32.4	29.3	32.7	32.8	33.6	29.5	28.6	32.1	31.6	31.4	32.1	30.8	32.4
32	31.2	32.8	29.1	32	31.7	33.1	29.2	28.9	31.3	32.6	32.2	31.9	30.9	32.1
30.7	29.8	30.5	28	31.1	31.3	32.5	28.2	28.2	30.1	31.3	30.9	30.8	30.6	31.2
32.5	31.7	32.5	30	32.9	33.5	34.4	29.7	29.5	31.8	32.5	31.6	32.3	31.1	32.5
36	31	34.8	38.3	33.4	37.2	35.1	35.2	35.5	35	36.4	37.2	36.3	35.2	37.6
33.5	31.1	31	33.7	30.1	34.8	32.3	31.3	33.5	33.7	36.9	32.3	33.7	34.1	35.6
36.1	31.1	35.5	35.6	33.6	39.3	36.9	34.3	35.2	35.2	37.9	36.8	37.7	35.1	37.5
35.8	35.2	33.6	37.5	35	37.4	34.7	34.5	36.6	36.3	39.2	35.6	36.9	38.2	39.3
36.3	31.9	37	38	34.7	38.7	35.8	35.4	36.1	35.8	37.3	38.1	38.1	36.4	38.8
38	34.4	35.5	37.5	33.6	39.2	36.5	34.8	37.2	38.4	39.6	37.3	38	38	40.1
34	30.6	31.9	33.7	30.3	37.2	34.5	31.8	32	33.3	36.7	33.9	34.3	34.1	36.4
36.3	32.1	34	34.8	31.7	38.8	36	33.4	35	34.5	37.1	36	36.8	37.2	38.2
32.8	31.7	34.1	35.5	33	35.6	32.6	33.4	33.4	34.6	34.7	34.1	34.6	33.4	33.5
37.4	34.9	34.9	36.8	32.9	39	36.1	34.2	36.8	37.7	40.2	37.2	37.3	37.8	40
33	30.5	33.3	37.7	32	36.8	33.6	34.3	34.9	33.7	35.6	35.6	34.3	35.3	35.1
37.4	31.2	36.2	34.8	34	38.7	37.2	35.4	35.7	35.8	37.7	36	39.5	35.8	37.7
36.9	31.2	34.7	39	33.9	37.8	36.1	34.7	36.1	35.7	38.1	37.7	38	36.8	37.8
36	29.8	34.2	35.1	32.3	39.1	35.6	33.5	33.8	34	36.1	35.5	37.2	34.4	35.9
36.3	31	34.2	39.5	34.1	38	36.9	36	36.7	38.3	38.9	38.2	38.5	37.6	39.2
38.4	34.1	36.2	41.3	35.9	40.1	38.3	36.9	37.8	37.8	40.1	40	39.9	39.5	41.2
35.7	33.3	34.2	40	33.5	39	36.7	37	37.4	35.9	38.1	37.7	37	37.7	36.6
34	31.5	35.5	38.8	33.8	38.2	34.2	35.7	34.8	34.9	35.5	36.8	36	36.1	35.6
33.9	30.9	30.8	32.2	30.8	35.9	33.7	30.8	32.2	32.5	35.7	32.9	33.1	33.8	35.7
35.2	31.9	33	33.2	30.4	36.8	34	32.4	34.2	35.4	36.5	33.6	34.7	34.4	37.1
35.5	31.5	33.1	33.9	31.3	38.2	35.9	32.1	31.3	32.9	35	34.5	35.1	33.8	35.6
35.1	31.3	31.6	34	31.3	36.5	33.6	32.5	35.3	35.3	38	34	34.9	36	37.9
35.4	31.4	36	37.1	33.3	38	35.2	34.5	34.9	35	36.8	37	37.6	35.3	37.1
32.5	31.3	35.3	36.2	32	35.5	32.8	34	34.4	34.8	35	35.3	34.8	34	33.4
32.8	32	32.3	37.1	33.6	35.4	34.4	34.3	34.4	35	34.8	34.7	33.8	35.1	34.5
35.1	33.6	32.1	34.3	33.1	38.2	35.5	32.8	31.8	34	36.4	34.6	34.5	35	35.1
35.2	30.5	32	32.8	29.6	37.3	33.8	31.6	31.6	32.6	36.1	34.5	34.6	34.4	37.2
35.5	31	35.3	35	34.3	37.8	34.3	33.6	31.4	35.1	34.5	33.8	35.8	33.2	33.4
34	29.1	33.3	36.1	31.9	35.9	33.3	33.1	33.8	33	35.3	36.6	36.4	33.4	35.3
37	33.8	35	36.7	31.9	37.7	35.8	32.8	35.8	36.4	37.9	36	36.6	36.3	38.6
37.3	31.5	35.8	36.5	33.2	39.7	36.8	34.2	34.7	34.2	36.8	36.2	37.5	34.3	36.8
34.4	32.3	33.9	39.1	33	37.3	36	35.9	36.5	35.4	37.7	37.1	36	37.4	36.4
38	32.6	35.8	35.2	33.5	39.8	37.7	34.4	34.6	35.1	39.5	36.5	37.5	35.9	38.2
34.3	32.5	32.1	34	32.7	37.2	35.4	32.4	31.9	33.4	36.7	34.5	34.2	34.5	35.4
35	31	34.1	37.5	32.7	36.9	34.9	34.5	35	35.1	36.2	37.7	36.9	35.2	37.1
33.9	30.5	34.8	37.2	32.4	37.1	33.1	35.2	34.4	33.3	35.6	36.1	34.3	33.4	35
36.5	31.7	36.4	38	33.5	38.2	35.1	34.9	35.2	36.4	37.2	36.7	38.2	35.4	37.4
34.3	29.4	33.6	36.9	31.9	36.2	33.5	33.5	33.6	33.4	34.8	35.4	35	33.2	35.5
35.8	31.2	36	35.6	34.1	38	36.3	34.9	33.2	35.1	35.3	34.8	38.5	33.8	34.1
34.2	30.4	35.3	36.1	33	37.2	33.6	34	33	34.2	35	34.9	35.1	33.6	34.6

35.1	31.1	36	35.2	34.1	38.1	36.2	34.1	32.9	35	35.5	34.8	37	33.8	34.6
33.8	30.9	35	37.5	33.1	37	33.3	34.1	33.2	34.3	35.2	34.7	34.2	33.8	34.1
38.9	32.4	37.8	35.7	36	41.3	37.6	33.7	35	36.6	37.6	36.5	38.9	34.9	36.5
37.3	32.1	35.5	34.7	34.6	39.8	36.7	33.8	33.5	36.1	37	34.7	37.5	34	35.3
37	30.4	35.9	34.6	33.5	39.5	35.5	33.3	32.5	33.7	33.8	34.4	37.5	32.9	33.6
37.6	32.7	35.5	36.8	33.1	40.4	37.9	34.4	33.8	35.1	37.3	37	37.3	36.1	38.5
36.2	34.7	33.8	40.2	34.2	36.6	37.9	37	37.3	37.5	38.8	38.5	37.7	37.4	37.1
36.2	31.6	35.8	35.6	33.7	38.7	35.7	33.5	33.1	34.4	35.7	34.9	37.9	33.6	34
35.4	30.5	34.2	33.2	33.4	37.4	34.5	32.4	30.8	33.9	34.7	33.2	35.2	31.9	33
29.2	28.4	28.2	27.7	26.9	24.9	25.5	27.4	27.4	24.6	27.9	27	28.4	27	27.6
32.2	33.9	31.8	30.5	31	30	31.6	31.9	31.8	29.2	31.9	32.4	32.9	31.5	31.8
35	35.2	33.3	31.9	32.2	31.3	33.4	32.6	33.3	31.5	33.8	34.2	35.3	33.5	33.5
33.2	34.2	34.4	34	33.3	32.2	32.5	34	34.2	31.7	32.9	33.8	34	33.8	34.1
35.2	36	34.8	33.4	34.2	33	34.5	35	35.3	31.6	33.9	35.5	35.8	33.8	34.3
34.4	35.5	34.3	34.8	34.5	33.2	33	34.6	34.1	31.5	35.2	34.4	35.2	34.1	35.3
34.6	35.7	34.3	34.1	33.7	31.6	33.5	34.9	34.1	31.4	34.4	34.1	35.9	34.3	35
32.9	33.8	32.2	30.7	31	30.3	32	32.6	31.9	28.7	31.7	32.9	33.2	31.7	30.9
35.6	36.1	34.9	33	33.1	32	34.3	34.7	33.8	30.8	33.5	34.4	35	33.2	33.4
38.9	38.9	38.4	39.5	39.6	37.2	37.6	38.5	38.6	36.8	40.1	39.8	40.3	39.1	39.7
28.4	31.1	29.6	29.2	29.3	27.5	27.5	28.9	28.5	26.4	30.3	29	30.4	28.7	29.9
34.8	34.6	33.7	33.2	33.1	31.9	34.7	33.4	33.4	31.1	33.9	34.5	35	33.6	33.8
35.3	36.3	35	36.2	34.9	35.4	34.8	35.6	36.2	32.2	36.1	35.3	36.4	35.5	36
35	36.4	34.3	33.1	33.3	32.7	34.4	34.8	34.5	31.3	34.2	35.1	35.5	33.8	33.9
31	32.4	30.2	28.2	29.1	27.8	30.2	30.3	29.6	26.9	29.5	30.4	30.6	29	29.2
35.1	35.8	34.8	34.4	34.6	33.1	34.1	34.9	35.1	33.2	35.6	35.7	36.4	35.7	37
32.4	31.9	32.2	31.4	32	29.9	30.5	31.5	31.7	28.7	31.5	32.2	32.5	31.4	31.9
33.8	34.2	33.4	34.3	32.9	31.5	31.9	33	33.1	31.3	33.9	34.2	34.5	33.2	33.4
37.6	38.5	37.3	38.1	38	36.8	36.5	37.5	37.2	35.1	38.4	37.7	38.1	37.5	38.3
32.3	31.9	30.5	31	30.9	28.6	30.4	31.2	31.9	28.6	31.7	32.8	33.3	33	32.9
31.8	32.6	31	32.3	30.8	30.8	30.7	31.5	32.3	28.5	31.9	32.1	32.9	31.9	31.6
34.4	35.3	33.6	32.4	32.7	31.9	33.6	34.3	33.8	30.4	33.4	34.2	34.6	33	33.1
36.6	37.2	36.4	37	36.3	36.2	35.3	35.9	36.3	33.8	35.9	36.4	37.6	36.6	36.5
34.2	34.7	34.2	33.2	33.2	32.1	32.5	34.1	33.9	31.4	34.4	34.5	34.8	33.6	34.4
32.8	33.8	32.4	32.9	32.5	30.6	30.7	32.3	32.5	29.9	32.8	33	33.4	31.7	32.9
31.6	32.4	31.2	31.5	31.5	30.2	29.9	30.8	31.6	29.2	30.2	31.3	33.2	30.7	31.6
37.9	39.2	38.1	38.1	37	36	36.5	37.9	37.7	34.7	37.6	38	39	37.8	38.1
37.6	39.1	37.8	38	36.7	35.4	35.9	37.6	37.6	34.3	37.4	37.5	37.9	37.2	38.8
25	25.7	24.6	23.6	23.5	21.9	22.6	24.4	24.1	21.9	24.4	24.4	25.3	24.1	24.8
33.7	34.4	33.2	32.2	32.3	31.3	33	33.2	33.3	30.9	33.1	33.3	34.4	32.6	33
33.2	33.8	32.9	31.9	32.3	31.1	31.6	32.9	32.9	30.6	33.2	33.6	33.8	31.9	32.8
34.4	35.9	35	36	34.2	32.5	33.6	34.4	34.2	31.5	33.8	34.4	35.6	34.4	35.1
28.4	29.6	28.9	29	28.3	27.3	27.8	29.2	29	27.6	29.8	29.2	29.7	28.9	29.3
34.1	35.2	33.7	34.1	33.9	31.6	32.2	33.8	33.4	30.9	34	34.3	34.7	33	34.5
28.1	28.9	28.5	25.4	25.6	25.5	27	26.8	27	23.9	26.9	26.9	27.5	26.5	26.4
34.3	34.7	33.3	32.8	33.1	31.5	32.3	33.5	33.5	31.1	33.7	34.1	35.1	33.4	34.5
32.6	33.8	32.8	33.1	32.5	30.8	31.8	32.3	32.9	30.6	32.8	33.4	33.6	31.6	31.9
31.7	32.7	31.8	31.6	31.6	29.1	30.1	32.1	31.3	29.2	31.1	31.2	32.5	31.3	31.6

32.6	33.1	31.8	31.9	32.1	29.3	30.9	32.8	32.3	30.3	33	33.2	33.5	31.7	35
31.7	32.7	32.1	32	31.6	30.2	30.7	31.8	31.9	29.4	31.6	32.1	32.5	31.1	31.5
30.3	31.1	29.5	28	33.1	32.2	31.1	29.7	28.6	28.6	34.5	30.1	30.1	31.5	32.2
34	32.7	33.4	31	35.5	35	35.6	32.6	32.3	32.9	34.2	34.4	33	34.6	35.7
30.8	30.2	31.2	28.2	33.6	32.5	33.9	29.8	30	29.7	32.2	32.5	30	31.9	33.1
27.7	27.5	28.1	25.7	30	29.8	29.6	27.1	27.4	28.1	29.2	28.8	28.2	29.8	30.5
31.6	31.1	30.6	29.9	33.5	32.3	31.9	30.9	29.7	30.5	32.2	31.5	29.5	31.8	33.9
30	29.1	29.2	27.6	32.3	31.9	31.6	28.5	27.9	28.5	30.2	30	29.1	31.2	31.4
26.2	25.3	26.2	23.2	27.1	25	27.7	25.6	24.7	25.5	27.2	26.7	26.3	26.8	27.5
33	32.9	32.3	31.1	34.9	34.4	34.5	32.1	31.4	31.6	33.7	33.1	31.8	34.6	35.4
31.1	30.9	31.5	29.4	33.4	31.8	32.7	30.9	30.2	30.6	32.1	32.5	30.4	32.3	33.4
30.1	29	29.1	26.9	32.4	31.6	31.7	29.6	28.3	28.7	30.8	30.6	29	31.4	31.8
30.8	29	30.2	27.5	31.8	32.1	32.7	29.5	28.8	30	31.3	31.2	30.4	31.1	32.4
30.9	30.2	30.5	29.3	33.7	32	32.3	29.4	30.4	29.3	31	32.4	30.2	31.2	32.8
31.7	30.8	32.2	28.8	33.3	32.8	34.7	29.7	30.3	30.2	31.7	32.7	30.6	32.5	33
30.9	30.8	31	28.3	33.1	32.8	32.5	29.8	29.4	29.6	31.4	32.2	30.7	32.2	33.6
30	28.4	30	27.7	32.5	31.8	31.4	28.5	28.9	29.8	31.2	31.3	30.2	30.8	32.1
33	31.8	32.7	31.4	34.3	34.1	34.7	31.1	31.4	31.9	33.3	33.3	32.6	33.1	34.2
30	29.2	30.8	27.3	32.3	32.2	31.6	28.9	28.5	29.2	30.5	30.6	29.7	30.6	31.5
27.2	26.6	27.6	24.4	29.3	30.1	29.6	26.5	26.5	26.3	27.7	29.5	28.2	28.8	30.2
30.5	29.2	30.2	27.7	31.6	31	31.8	29	28.3	29.7	31.6	30.8	30.6	30.9	32.5
28.3	25.6	27.7	24.9	29.1	29.1	29.3	28.3	26.8	27.6	28	26.2	30.6	29.5	30.1
24.8	23.1	24.8	22.8	25.5	25.2	25.5	24.4	24.2	24.1	25.1	23.9	26.5	25.5	26.9
29.1	26.5	27.5	25.4	29.3	29.3	29.2	29.2	27.7	27.7	28.7	26.5	29.8	29.6	30.1
27.6	25.4	26.3	24.9	28.8	28.4	28.4	28.1	26.9	26.7	27.9	25.3	29.8	28.7	29.7
27.8	24.8	26.4	24	28.1	27.8	27.7	27.6	26.6	26.2	27.2	25.7	28.6	28.1	28.6
27.5	24.7	25.9	24.4	27.9	27.7	27.7	27	25.5	25.9	26.6	24.9	28.9	28.3	28.6
29.1	26.8	27.5	25.9	29.4	29.4	29.3	28.9	27.6	27.7	28.5	26.5	30.8	29.9	30.7
27.3	24.5	25.8	23.8	27.4	27.7	26.9	27.1	26.1	26.1	27.5	26	28.6	28.4	29
32	33.9	29.5	31.2	29.6	27.7	30.6	32.1	32.5	32	27.4	32.6	30.8	28.3	28.9
30.5	30.3	28.3	29	27.4	25.8	28.8	29.8	31	29.1	26.6	31.1	30.1	26.9	28
33.5	33	31.1	32.8	30.5	28	31.2	33.2	33.7	33	30	33.1	32.9	29.9	32.5
25.9	25.3	23.8	26.4	23.3	20.9	23.1	25.4	26.3	26.5	21.5	26.1	27.6	22.9	23.3
25.9	25.7	25.1	28.3	24.7	23	24.1	26.3	27	28.5	24.3	26.4	28.6	24.8	25.2
32.7	34.1	28.9	29.7	28.8	27.8	29.1	31.4	31.1	32	28.8	33.4	31	28.3	29.1
33.3	33.3	31	33.4	31.2	29	31.4	33.6	34.3	33.1	29.2	34.1	33.6	29.9	31.8
35.2	34.3	33	34.4	32.4	30.1	32.8	35	35.4	34.8	31.7	34.8	34.6	31.7	33.1
32.2	32.2	29.9	31.6	29.5	27.7	31.1	32.5	32.7	31.8	28.6	32.1	31.6	28.7	30.2
31	32.1	28.8	29.7	28	26.5	29.6	31.2	31.9	30.6	27.2	31.4	30.7	27.5	29
31.4	31.3	28.6	29	28	25.8	28.4	31	31.5	31.1	26.9	31.5	29.9	27.7	27.9
31.5	30.9	29.5	30.2	29	25.3	29.6	29.6	31.4	31.8	27.4	31.3	29.8	26.6	28.6
27.6	27	27	29.8	25.7	25	25.3	27.3	28.3	28.4	25.7	27.9	29.2	26.1	26.7
31.2	31.4	29.7	32	29.5	26.6	30	32.5	32.1	31.6	28.6	31.6	31.9	28	29.3
22.8	22.4	21	24.2	21.6	20.5	21	22.6	24	25.1	22.5	22.8	25	22.5	22.1
30.1	29.9	29	30.1	28.6	24.5	28.7	29.3	29.9	29.2	26.4	30.7	30.1	26.8	27.9
25.9	24.7	23.9	27.5	24.3	22.3	23.1	25.1	26.6	27.6	23.2	25.8	27.5	23.5	24.1
22.2	21.9	20.4	23.2	21.2	20.5	20.3	22.5	23.4	23.3	22.3	22.8	24	20.9	21.8

30.5	31.1	28.4	28.9	27.3	24.8	28	30.3	30.9	30	26.7	30.3	29.5	26.7	27.8
31.1	31.4	29.1	30.7	29.3	26.6	29.8	31.4	31.6	31.2	27.2	31.3	31	27.7	29.4
32.2	33	29.9	31.4	29.8	26.3	30.2	30.6	31.6	30.8	27.1	30.9	30.4	27.5	28.9
32.9	32.4	30.6	31.8	31	28.5	30.6	32.9	33.3	32	29.1	32.7	32.2	28.7	29.9
28	28.7	26.4	29.1	27.1	23.2	27.3	29.3	29.2	29.3	24.8	28.9	28.8	25.5	27.2
31.5	31.6	29.5	30.8	29.4	25.4	29.3	30.9	31.6	32.1	27.9	31.5	30.5	27.6	28.5
25.7	25.1	24.1	26.9	23.9	22.9	22.7	26.1	26.7	27.7	24.1	25.9	27.5	24.5	24.5
23.6	23.3	22.6	25.8	23.4	21.5	21.9	25.3	24.9	26.1	23.3	24.6	25.4	23.6	22.6
24.2	23.8	22.6	25.3	22.8	21.3	21.5	24	25	26.4	22.9	24.5	26	23	23.6
22.6	22	20.3	23.8	20.1	18.2	19.3	21.7	22.4	23.6	19	21.5	23.1	19.6	19.7
25.4	24.3	23.7	27.1	23.6	21.4	22.4	25	26.3	27.4	23.5	25.5	27.2	24.1	24.1
26.1	24.8	24.9	27.6	24.3	23.5	23.1	24.8	26.6	27.9	24	25.5	27.6	24.4	24.5
24.6	24.5	23.1	26.4	23.8	22	22.8	24.8	25.1	26.2	23.5	24.4	26.8	23.9	23.2
20.5	21	18.8	22.8	20.5	19.3	19.6	19.5	22.4	24.1	20.7	21.2	23.1	20.5	20.4
29.4	30	27.6	29.9	27.4	23.8	29	29.9	30.5	29.6	26.5	29	29.6	26.2	28
32.6	33.6	30.2	31.4	29.4	27.7	30.5	32.8	32.6	31.5	28.5	32.3	31.5	28.5	29.9
33.7	33	31.8	33.2	31.4	28.2	33.1	33.8	33.4	33.2	29.8	33.4	32.9	29.5	31.1
32	30.9	29.2	30.3	28.8	26.3	29	32.2	32.6	30.7	27.9	31.6	31.1	28.1	29.1
22.8	22.4	20.9	24	22	20	20.2	21.8	23.3	24.5	20.6	21.7	23.7	21.9	22
19.2	19.5	17.9	20.6	18.3	17.4	17.8	18.9	20.5	21.2	18.1	19.4	20.5	19.1	19
22.5	22	21.6	22.2	21.4	20.4	21	23.4	22.5	23.9	22.4	23.4	24.1	21.8	21.2
25	24	23.3	25.9	23.1	22	21.7	24.9	25.9	27.2	23.4	24.9	26.9	23.6	24.3
24.6	24.3	23	26.7	23.7	20.5	22.4	25	25	25.9	23.7	23.9	26.4	23.5	23.8
18.3	19.1	17	18.5	16.8	17.9	17.4	17.9	18.7	20.9	18.8	18.7	18.7	19.5	18.5
22.3	22.3	20.8	23.5	19.3	20.4	19.9	20.6	22.8	24.1	20.7	21.9	22.6	21.1	20.8
20.6	22.1	18.8	20.6	17.8	19.2	19	19.6	20	21.9	20	20.1	20.1	20.8	19.6
29.7	27.6	27.1	23.9	27.2	28.7	30.5	26.8	24.7	26.9	27.2	26.9	27.9	27.9	28.5
30.5	27.6	26.4	23.6	27.3	28.3	29.5	26	25.2	26.9	26.3	26.1	27	27.9	28.2
29.4	27.8	27.1	23.8	27.5	29.3	32.1	27.4	25	26.9	27.7	27	28.5	28	28.9
27.7	25.8	25	21	24.6	26.4	27.3	24.6	22	24.2	23.8	24.1	24.6	26	25.5
29.2	27.2	27.1	24.6	27	28.3	29.6	26	25.1	27	27.1	26.7	27.9	28.1	29.4
29.5	28.8	28.3	24.7	30.1	31	33.8	28.3	26.3	27.3	28.3	28	28.7	30.2	29.8
29.7	27.7	27.3	25.3	29.6	30.6	32.4	27.2	25.6	27	28.6	27.4	28.6	29.4	29.7
27.2	26	26.7	23.6	27.8	28.2	29.8	25	24.2	24.9	26.8	25.6	25.9	27.2	28
29.8	26.8	25.8	22.6	26.2	27.6	29.5	25.9	23.5	25.4	25.6	25.8	26.9	26.6	28.2
29.6	28.5	28.2	24.1	29.3	30.3	33.8	27.6	25.8	26.9	27.6	27.6	28.3	29.5	29.3
30.2	28.2	27.6	25.4	28.9	30.3	32.1	27.3	25.5	27.3	28.2	27.5	28.9	29	29.5
29.2	28.7	27.7	24	28.4	29.8	32.8	27.6	25.1	27	27.7	27.2	28.2	29	29
31.2	28	27.6	24.2	27.4	28.9	29.8	26.6	25.2	26.3	26.8	26.4	27.2	27.6	27.8
29.2	26.5	26.3	23.7	26.9	28.4	29.9	26	24.3	25.9	26.4	26.6	27.5	27.5	27.5
29.8	27	27.2	24.3	28	29.2	31.6	26.2	24.7	26.3	27.8	26.6	28.1	28.1	28.6
30.5	27.7	26.9	23.9	27.3	28.8	30.6	26.5	25.1	26.5	27	27	27.7	27.7	28.5
29.9	27.9	27.4	24.6	27.6	28.7	31	26.9	25.7	26.8	27.9	27.4	28.9	28.9	29.1
29.2	28.9	27.7	23.2	27.5	28.4	30.9	27	24.8	26.5	27	27.1	28.5	27.5	28
28.4	27.3	27	22.3	28.3	29.8	32.6	27.2	24.8	26	26.7	27	27.5	28.3	28.4
28.4	27.6	27.3	24.3	27.9	30	32.3	26.8	25	26.2	27.5	26.8	27.7	28.8	29
26.8	25.5	25	23	25.5	26.2	27.1	23.3	22.5	23.4	25.1	23	23.7	24.7	25.4

28.4	27.3	26.7	23.1	26.8	28.1	30.6	26.1	23.9	25.5	26.2	26.3	26.6	27.5	27.7
29.7	28.5	28.5	24.4	28.6	31.1	32.8	28.5	26.3	27.6	28.2	28.4	28.8	29.8	29.9
30	30.6	29.7	27.9	33	33	31.1	29.5	28.9	28.8	31.1	30.2	29.4	31	32.1
28.4	27.5	28.1	25.2	29.9	30.6	30.6	27.9	27	27.7	28.9	28.6	28.3	29.2	30.5
31.5	31.3	31.7	28.2	32.5	32.5	33.3	31.2	29.6	30.9	32.2	31.6	32.9	32.7	33.3
29	27.4	28.7	25.7	30.3	30.8	31.4	28.4	26.2	28.3	29.3	28.8	29.5	30.7	31.8
27.9	26.8	27.7	25.4	29.7	29	30.2	28.2	27	27.9	28.6	30	28.1	30.3	30.3
25.1	24.1	24.1	22.1	26.2	26.8	27	24.8	23.6	24.5	25.5	25.2	25.6	26.1	27.5
29.3	28.3	28.9	26.7	30.2	31.3	31	28.9	27.5	28.9	29.5	29.4	30	30.8	32.1
28.7	28.1	28.3	26.6	30	31.2	31.1	29.1	27.7	28.4	29.3	28.7	30.4	30.5	32
29.5	28.5	29.1	26.8	30.6	31.5	31.7	28.8	27.4	28.7	30.3	29.2	30	30.9	32.7
28.9	27.8	28.6	26.5	30.3	31	31.6	28.2	27.4	28.4	29.9	28.8	29.7	30	31.7
28.5	28	28	25.7	29.7	30	29.7	27.9	27	27.6	29.2	28.7	29	29.3	30.2
28.2	27.3	27.5	25	29.2	30.3	29.5	28	26.2	27.4	28.4	27.8	28.9	29.6	31.2
27.9	30.4	29.4	27.1	31.4	32	32.8	29.6	29	29.6	30.5	30.7	30.3	31.2	33
29.6	30.1	29	26.2	27.5	30.9	31.6	29.8	30.5	27.5	28.8	31.1	28.7	27.5	28.6
30.9	32	30.5	28.2	28.9	28.8	30.5	30.8	30.2	26.7	29.8	30.9	30.9	29.6	30.2
24.2	25.4	22.7	20	21.3	22	23.9	24	22.5	20.4	21.5	23.8	24.2	22.3	21.5
35.2	35.5	33.5	31.4	32.1	34.9	35.1	33.8	34.4	31.4	34.1	35.6	33.8	32.6	34.2
29.8	30.9	28.4	26.2	25.4	26.9	28.4	28.3	27.2	24.9	27.3	30	28.7	27	26.7
32.5	33.6	32	30.1	30	29.8	31.7	32.2	31.5	28.9	31.3	31.8	31.7	31	31
26.8	28	24.4	22.6	23.3	24.2	25.8	25.3	24.7	21.6	23.9	25.4	25.4	23.7	22.6
28.1	29.3	27.2	25.2	25.6	25.6	27.2	27.1	26.7	23.4	26.7	27.5	27.6	25.8	26.5
28.7	28.9	27.7	26	23.1	25.6	28.4	28.5	28.2	25.6	27.2	28.5	28.8	27.2	27.3
28	30	26.5	24.4	24.8	25.8	27.6	27.7	27.2	24.3	26.2	27.8	27.1	25.9	25.4
30.1	30.5	31.6	26.5	28.7	29.8	30.5	28.9	27.6	26.9	30.1	30.3	28.6	29	30.3
34.8	34.7	30.6	29.6	29.3	31.1	34.9	31.8	30.4	29.6	32.1	33.8	30.6	29.3	32.7
33	33	30	28.1	29.2	31.5	32.2	30.5	30.4	28.8	29.9	32.2	29.7	28.1	29.4
28.2	29.1	26.8	24.3	25.1	26.3	26.7	26.6	26.4	23.3	27.3	27.4	26.5	25.9	26
29.5	30	29.5	27	28.5	29.4	30	28.9	28.1	27	29.6	30	29.5	29.3	30
35.6	36.4	34	31	32.9	35.1	36	34.3	33.6	32.2	33.3	34.6	33	32.2	33.7
29.3	29.6	28.9	26.7	27.4	28	29.4	28.5	27.9	25.8	28.6	29.2	29.4	28.4	29
33.2	31.9	31.4	28.5	29.9	31.2	32.4	31.6	31.3	29.3	31.3	32.5	31.5	30	32.4
34.3	34	33.3	29.3	30.8	33	35	30.9	31.2	30.6	32.3	33.8	31.2	31.3	33.4
32.6	31.9	32.2	28.1	29.9	31.2	32.6	30.2	29.9	29.7	31.6	32.5	31	30.7	31.9
34.5	33.1	29.4	26.8	28.4	30.5	33.5	30.3	28.5	29.1	31.3	33.5	28.7	27.8	30.8
32	31.5	30.6	27.1	28	30.4	32.5	28.7	28.7	28.1	30.1	31.1	28.4	28.1	30.5
33.4	33	31.9	29.2	30.2	31.5	33.1	32.2	30.4	28.9	31.7	32.1	31.4	30.4	31.5
33.3	32.5	31.9	29.2	29.8	32.5	34.9	32.2	31.1	29.6	31.5	32.9	31.2	28.9	32.1
32.3	32	32.3	28.2	30.3	31.6	33.4	29.2	29.4	29.3	31	32.4	30.5	30.3	31.9
31	30.9	29.4	26.5	28.6	30.2	30.7	29.7	28.5	27.2	30	29.9	29.4	28.8	30.5
33.2	31.9	32.1	28.8	29.4	31.7	34	30	30.1	29.9	31.9	31.9	30.2	29.1	31.6
29.9	29.7	28.7	25.7	27.7	28.7	30.1	29.8	27.5	26.4	28	29	28.3	27.8	28.3
33.9	33.1	30.5	28.3	28.4	31.9	33.6	30	29.5	27.6	30.2	32.1	28.8	29.2	31.2

2021	2022	2023	2024	2025				
32.3	34.8	33.3	33.8	34.7	trend=	0.17	AL	MUSCLE SHOALS AP
31.9	33.4	33.7	33.4	34	trend=	0.14	AL	TALLADEGA
31.6	34.1	32.7	33.1	33.3	trend=	0.12	AL	GAINESVILLE LOCK
32.4	33	34.5	33.6	34.3	trend=	0.11	AL	BREWTON 3 SSE
31.7	31.8	34.5	32.8	33.6	trend=	0.08	AL	FAIRHOPE 2 NE
31.9	33	33.9	33.9	33.7	trend=	0.08	AL	TROY
32	34	33.1	33.2	34.6	trend=	0.08	AL	TUSCALOOSA ACFD
33	34.8	35	35.4	35	trend=	0.07	AL	SELMA
30.9	32.8	31.8	31.8	32.6	trend=	0.02	AL	SCOTTSBORO
31.5	32.8	33.5	33.1	33.7	trend=	0.01	AL	THOMASVILLE
31.8	33.7	34.1	33.8	34.1	trend=	-0.01	AL	GREENSBORO
30.8	32.3	32.5	32.8	32.3	trend=	-0.05	AL	HIGHLAND HOME
28.9	30.7	29.5	29.9	31	trend=	-0.06	AL	VALLEY HEAD
30.1	32.7	31.2	31.9	32.4	trend=	-0.08	AL	SAINT BERNARD
31.6	33.1	33.2	33.3	33.5	trend=	-0.26	AL	UNION SPRINGS 9 S
31.6	36.8	33	32.3	33.1	trend=	0.14	AR	GRAVETTE
32.8	36.1	33.8	32.2	34.7	trend=	0.08	AR	CONWAY
30.1	34.8	32.3	31.5	31.9	trend=	0.05	AR	FAYETTEVILLE EXP STN
30.6	36.3	31.5	31.2	32	trend=	-0.02	AR	MENA
31.7	36.4	33.3	32.1	33.1	trend=	-0.02	AR	OZARK 2
31.5	36.6	33.8	32.2	33.7	trend=	-0.03	AR	SUBIACO
32.5	36.4	33	31.9	33.8	trend=	-0.10	AR	PRESCOTT 2 NNW
32.4	35	33.1	32	34.3	trend=	-0.12	AR	PINE BLUFF
31.1	35.3	32.5	30.3	33	trend=	-0.17	AR	MAMMOTH SPRING
31.5	34.3	32.2	31.6	33.5	trend=	-0.19	AR	BRINKLEY
30.3	35.1	32.1	30.7	31.8	trend=	-0.24	AR	EUREKA SPRINGS 3 WNW
32	34.7	32.5	31.5	34.3	trend=	-0.26	AR	CORNING
30.8	34.2	32	30.9	32.7	trend=	-0.26	AR	NEWPORT
31.8	34.5	31.9	30.8	33.9	trend=	-0.28	AR	POCAHONTAS 1
31.1	33	32	31.2	32.9	trend=	-0.28	AR	ROHWER 2 NNE
33.1	33.8	36.4	34.7	34	trend=	0.84	AZ	CANYON DE CHELLY
32.5	32.9	36.4	34.5	33.5	trend=	0.67	AZ	SAINT JOHNS
37.1	39.3	42.9	41	38.9	trend=	0.59	AZ	ROOSEVELT 1 WNW
32	34.1	38.1	35.3	33.6	trend=	0.57	AZ	PEARCE SUNSITES
38.5	40.4	43.3	42.3	40.1	trend=	0.56	AZ	WICKENBURG
29.2	29.1	32.1	31.8	29.4	trend=	0.55	AZ	GRAND CANYON NP 2
34	33.8	36.5	37	33.5	trend=	0.50	AZ	SELIGMAN
35.5	38.2	41.6	39.7	38.4	trend=	0.49	AZ	TUCSON WFO
26.7	26.9	30.4	30.9	29.1	trend=	0.44	AZ	FT VALLEY
38.6	40.5	43.5	42.4	40.3	trend=	0.43	AZ	CHANDLER HEIGHTS
39.2	38.5	40.7	40.6	39.5	trend=	0.43	AZ	LEES FERRY
31	32	35.5	34.9	33	trend=	0.42	AZ	PRESCOTT
28.1	27.9	31.5	31.2	29	trend=	0.41	AZ	WILLIAMS
41.9	42.3	44.3	44.4	41.5	trend=	0.40	AZ	YUMA CITRUS STN
43.2	42.2	45.5	46.8	42.6	trend=	0.37	AZ	PARKER
31.8	33.5	37.6	35.1	34.1	trend=	0.37	AZ	TOMBSTONE

39.7	40.7	44.6	43.6	41.1 trend=	0.36 AZ	AJO
34.8	35	38	36.5	35 trend=	0.36 AZ	HOLBROOK
41	42.8	45.8	44.4	42.4 trend=	0.33 AZ	BUCKEYE
37.7	39.3	42.7	41.6	39.1 trend=	0.31 AZ	CHILDS
35.2	35	38.5	39.1	36.3 trend=	0.31 AZ	KINGMAN #2
36	38.4	41.4	39.2	37.4 trend=	0.30 AZ	SAFFORD AGRICULTRL CTR
39.1	41.1	44	43.3	41.3 trend=	0.16 AZ	SACATON
32.1	33	36.2	34.9	33.3 trend=	0.14 AZ	WHITERIVER 1 SW
36.1	38.2	41.5	40	37.9 trend=	0.12 AZ	MIAMI
38	35.2	35.7	38.4	35.3 trend=	0.94 CA	ORLEANS
37.5	35.5	36.5	38	34.9 trend=	0.79 CA	YREKA
34.9	32	32.9	35.2	31.2 trend=	0.78 CA	COLFAX
35.8	34.8	34.1	35.4	33 trend=	0.74 CA	CEDARVILLE
38	36.4	36.9	38.9	36 trend=	0.73 CA	HAPPY CAMP RS
35.7	34.6	34.9	37.1	30.9 trend=	0.69 CA	FAIRMONT
35.3	32.9	34	36.2	32.2 trend=	0.68 CA	QUINCY
29.2	27.1	27.4	27.5	25.4 trend=	0.66 CA	TAHOE CITY
38	35.6	36.6	38.8	35.1 trend=	0.66 CA	WEAVERVILLE
20.2	19	20	21.3	18.1 trend=	0.65 CA	FT BRAGG 5 N
40.4	39.9	40.8	41.6	38.3 trend=	0.63 CA	INDEPENDENCE
32.7	32.3	33.5	33.6	29.7 trend=	0.60 CA	OJAI
38.4	36.8	38.4	40.8	35.7 trend=	0.58 CA	MARYSVILLE
43.4	43.4	45.4	45.8	42.6 trend=	0.57 CA	BRAWLEY 2 SW
30.2	29.2	30.5	32.3	28.2 trend=	0.57 CA	CUYAMACA
33.2	32.8	34.3	34.4	31 trend=	0.57 CA	PASADENA
39.7	38.5	39.2	41.5	36.6 trend=	0.57 CA	REDDING MUNI AP
37.4	36	37	39.4	34.5 trend=	0.56 CA	VACAVILLE
39.5	37.5	38.1	40.1	36.7 trend=	0.55 CA	HANFORD 1 S
41.7	41.9	43.7	44.4	41.6 trend=	0.55 CA	INDIO FIRE STN
21.8	21.4	22	21.5	21.2 trend=	0.54 CA	NEWPORT BEACH HARBOR
40.2	38	39	39.8	35.6 trend=	0.54 CA	WASCO
25.3	24.7	25.4	24.7	23.8 trend=	0.53 CA	SANTA BARBARA
38.7	37.2	36.4	38.7	34.3 trend=	0.53 CA	WILLOWS 6 W
34.5	33.7	36.2	37.2	30.5 trend=	0.52 CA	PASO ROBLES
16.1	17.5	17.6	18	16.5 trend=	0.51 CA	EUREKA WFO WOODLEY IS
31	28.8	29.6	31.8	27 trend=	0.50 CA	MT SHASTA
39.6	38.2	38.8	40.7	36.1 trend=	0.49 CA	FRESNO YOSEMITE AP
36.8	35.1	36.2	38.1	34.2 trend=	0.49 CA	MERCED
26	25.3	27	28	23.9 trend=	0.48 CA	SAN LUIS OBISPO POLY
37.4	36.1	36.5	39.1	34.8 trend=	0.48 CA	TEJON RANCHO
48.2	47.5	49.2	50.8	46.7 trend=	0.47 CA	DEATH VALLEY
40.2	38.5	39.3	40.9	37.2 trend=	0.47 CA	LEMON COVE
37.5	36.2	37	38.8	34.5 trend=	0.46 CA	ELECTRA P H
34.4	32.9	34.3	36.4	31.6 trend=	0.45 CA	LODI
37.7	36.3	38	38.3	34.2 trend=	0.43 CA	REDLANDS
37.4	36.2	36.1	38.8	34.6 trend=	0.42 CA	CHICO UNIV FARM
43.6	43.6	46	46.2	43.6 trend=	0.41 CA	BLYTHE

43.7	43.2	46	46.8	43.3 trend=	0.40 CA	NEEDLES AP
34.8	33.1	33.4	34.8	30.8 trend=	0.40 CA	YOSEMITE PARK HQ
29.6	27.9	28.8	31	26.8 trend=	0.38 CA	LAKE SPAULDING
37.1	36	35.2	38.1	33.3 trend=	0.36 CA	ORLAND
34.4	31.5	34	36.3	30.9 trend=	0.32 CA	LIVERMORE
30.4	29.4	31.6	33	28.4 trend=	0.31 CA	HEALDSBURG
24.7	23	23.6	24	22.7 trend=	0.29 CA	CHULA VISTA
29.1	28	30.7	30.1	24.9 trend=	0.26 CA	SANTA ROSA
31.4	30.7	32	32	30 trend=	0.26 CA	TUSTIN IRVINE RCH
36	34.8	35.9	38.3	33.3 trend=	0.21 CA	DAVIS 2 WSW EXP
36.2	32.7	34.4	36.9	31.9 trend=	0.17 CA	UKIAH
33.7	32	32.5	34.7	30.9 trend=	0.15 CA	SUSANVILLE 2SW
26.5	26.6	28.2	31.9	25.4 trend=	0.14 CA	NAPA STATE HOSPITAL
19	22.7	23.6	25.7	21.5 trend=	0.11 CA	BERKELEY
23.4	23.7	23.2	24.5	21.9 trend=	0.10 CA	SANTA CRUZ
26.5	25.9	28.4	29.4	23.9 trend=	-0.02 CA	PETALUMA AP
27.7	26.8	30	27.5	27.2 trend=	0.71 CO	DEL NORTE 2E
33.1	31.8	34	32.6	32.7 trend=	0.57 CO	COLLBRAN
33.6	33.1	35.3	33.4	33.5 trend=	0.55 CO	MONTROSE #2
28.2	26.8	28.5	26.5	27.6 trend=	0.53 CO	GUNNISON 3SW
36	36.4	36.8	36.7	36.2 trend=	0.52 CO	FRUITA
25.2	24.3	25.3	23.7	24.3 trend=	0.51 CO	DILLON 1 E
28.8	27.7	30.5	27.5	28.3 trend=	0.47 CO	SAGUACHE
31.5	31.9	30.1	31.9	32.2 trend=	0.45 CO	FT COLLINS
25.6	24.1	27.4	25.6	25.8 trend=	0.42 CO	TELLURIDE 4WNW
31.2	32.3	32.5	30.1	31.7 trend=	0.42 CO	TRINIDAD
32.7	34	30.7	32.5	32.9 trend=	0.40 CO	WRAY
30.6	32	30.4	31.7	31.4 trend=	0.35 CO	BOULDER
31.1	32.3	30.1	31.2	31.6 trend=	0.33 CO	CHEYENNE WELLS
27	26.9	29.5	26.8	27.1 trend=	0.32 CO	MANASSA
30.3	28.2	28.8	28.1	29.5 trend=	0.32 CO	STEAMBOAT SPRINGS
29.4	29.1	29.1	29.2	29.4 trend=	0.28 CO	CHEESMAN
32	32.5	31.8	32.1	32.8 trend=	0.25 CO	ROCKY FORD 2 SE
33.8	34.5	32.8	33.5	34.5 trend=	0.15 CO	LAMAR
32	34.2	31.9	33.6	32.6 trend=	0.11 CO	EADS
33.7	34.5	30.9	33.4	33.8 trend=	0.11 CO	FT MORGAN
25.4	25.1	27.9	24.8	25.2 trend=	0.10 CO	HERMIT 7 ESE
30.8	32.7	32.5	31.5	31.9 trend=	0.09 CO	CANON CITY
34.5	35.8	34.2	35.2	35.4 trend=	0.09 CO	LAS ANIMAS
33	34.6	31.8	33.5	32.8 trend=	-0.15 CO	HOLLY
26.9	30.4	29.6	29.5	30.1 trend=	0.47 CT	FALLS VILLAGE
27.9	30.3	29.6	29.6	30.5 trend=	0.42 CT	STAMFORD 5 N
25.3	27.5	27.7	27	28.2 trend=	0.22 CT	STORRS
25.9	28.2	27.6	27.1	28 trend=	0.10 CT	GROTON
29.8	30.2	29.5	30.3	30.5 trend=	0.52 DE	WILMINGTON PORTER RES
31.4	32.4	32.1	32.8	32.6 trend=	0.49 DE	NEWARK UNIV FARM
30.2	30.9	30.1	31.1	31.3 trend=	0.38 DE	DOVER

31.8	31.9	32	31.6	32.3	trend=	0.34	DE	MILFORD 2 SE
30.7	30.9	31.1	31	31.4	trend=	0.30	DE	GREENWOOD 2NE
32.7	33.7	34.5	33.8	33	trend=	0.32	FL	EVERGLADES
33.4	33.3	34.7	34.6	34.6	trend=	0.32	FL	TALLAHASSEE WSO AP
31.8	32.6	33.6	32.4	32.7	trend=	0.30	FL	FT PIERCE
33.5	34.5	35	34.8	34.3	trend=	0.28	FL	BELLE GLADE
31.3	31.8	33.3	32.7	31.9	trend=	0.28	FL	KEY WEST INTL AP
32	33.4	33	33.5	32.9	trend=	0.25	FL	SAINT LEO
31.1	31.7	32.7	32.5	32.3	trend=	0.23	FL	APALACHICOLA AP
32	32.7	33.9	33.3	32.9	trend=	0.23	FL	FT LAUDERDALE
31.9	32.5	33.2	33.2	32.8	trend=	0.22	FL	TITUSVILLE
33	33.3	34.6	34.1	34	trend=	0.19	FL	DE FUNIAK SPRINGS 1
33.3	34.2	34.5	34.1	33.9	trend=	0.19	FL	FT MYERS PAGE FLD
31.5	32.8	34.2	34	35.1	trend=	0.17	FL	LAKE CITY 2 E
32.6	34.2	33.1	33.9	33.5	trend=	0.17	FL	TARPON SPGS SEWAGE PL
31.8	32.5	33.3	33.1	32.8	trend=	0.14	FL	PERRINE 4W
31.1	32.6	32.8	32.6	32.5	trend=	0.09	FL	FEDERAL POINT
32.3	34	34.2	34.1	34	trend=	0.08	FL	BARTOW
32	33.4	34	33.7	34.2	trend=	0.03	FL	FERNANDINA BEACH
32.2	32.9	33.8	33.8	34.5	trend=	0.02	FL	MADISON
31.2	31.6	33.3	32.2	32.9	trend=	0.00	FL	PENSACOLA RGNL AP
32.5	33.9	34.3	34.2	33.4	trend=	-0.02	FL	ARCADIA
32.6	33.7	34.1	34.4	34.1	trend=	-0.04	FL	INVERNESS 3 SE
32.3	33.6	33.3	33.6	33.7	trend=	-0.10	FL	OCALA
29.9	31.1	31.3	31.3	32.6	trend=	0.31	GA	GAINESVILLE
32.5	33.7	34.3	34.4	35.3	trend=	0.28	GA	MILLEDGEVILLE
32.4	33.8	34.3	33.9	34.2	trend=	0.24	GA	SAVANNAH INTL AP
31.4	33.5	32.4	32.4	34.7	trend=	0.21	GA	ROME
31.9	32.7	33.4	33.5	34.1	trend=	0.21	GA	TIFTON
32.4	33.3	34.3	33.8	35.1	trend=	0.20	GA	HAWKINSVILLE
32.4	33.8	34.1	33.9	34.4	trend=	0.15	GA	BRUNSWICK
31.9	32.9	33.3	33.9	34.3	trend=	0.15	GA	MILLEN 4 N
33.5	34.9	35.7	35.4	36.3	trend=	0.15	GA	WAYCROSS 4 NE
33	34.2	35	34.9	35.5	trend=	0.14	GA	GLENNVILLE 3NW
31.4	32.6	33.5	33.6	34	trend=	0.13	GA	BAINBRIDGE INTL PAPER
32	33.2	33.8	33.6	34.7	trend=	0.13	GA	QUITMAN 2 NW
31	32.3	32.3	32.6	34.4	trend=	0.12	GA	TOCCOA
31.1	32.1	32.9	32.6	33.7	trend=	0.11	GA	COVINGTON
32	32.8	34	33.3	34.5	trend=	0.09	GA	EASTMAN 1 W
32.1	33.1	33.9	34	34.9	trend=	0.08	GA	ALBANY 3 SE
29.8	30.5	31.4	31.5	32.2	trend=	0.06	GA	NEWNAN 5N
31.7	33.1	33.2	33.7	34.7	trend=	0.06	GA	WARRENTON
30.2	31.8	31.3	31.3	32.5	trend=	0.04	GA	DAHLONEGA
31.3	32.7	32.9	32.9	33.6	trend=	0.04	GA	WEST POINT
30.5	31.2	32.2	32.2	32.9	trend=	0.03	GA	CAMILLA 3SE
30.9	31.8	32.7	32.8	33.2	trend=	0.03	GA	TALBOTTON
30.8	32.2	32.2	33	34	trend=	0.03	GA	WASHINGTON 2 ESE

28.2	29.1	27.4	27.8	28.6	trend=	0.00	IA	STORM LAKE 2 E
28.6	28.8	27.9	27.7	28.3	trend=	-0.01	IA	ESTHERVILLE 2 N
28.5	29.8	29	28.3	29.7	trend=	-0.06	IA	MT PLEASANT 1 SSW
28.5	30	29	27.8	29.8	trend=	-0.11	IA	BELLE PLAINE
27.5	27.8	27.2	26.4	28.1	trend=	-0.13	IA	NEW HAMPTON
29.5	30.7	29.3	28.7	29.9	trend=	-0.14	IA	INDIANOLA 2W
29.3	30.4	28.9	28.9	29.3	trend=	-0.15	IA	MT AYR
29.2	29.5	28.2	28.1	29.1	trend=	-0.18	IA	ALGONA 3 W
27.2	27	26.5	25.9	27.2	trend=	-0.19	IA	FAYETTE
28.6	28.9	28.4	27.2	29	trend=	-0.21	IA	TOLEDO 3N
29.2	30.4	30.2	28	30.3	trend=	-0.21	IA	WASHINGTON
29.5	30	27.8	28.3	29	trend=	-0.22	IA	ROCK RAPIDS
27.7	28	27.2	26.9	27.9	trend=	-0.23	IA	FOREST CITY 2 NNE
28.2	29.3	28.5	27.3	29.1	trend=	-0.25	IA	FAIRFIELD
29.6	30.2	28.6	28.4	29.5	trend=	-0.25	IA	LE MARS
27.6	27.8	27.4	26.5	27.4	trend=	-0.27	IA	CHARLES CITY
27.9	28.7	28.5	27.3	29.7	trend=	-0.28	IA	CLINTON #1
28.1	28.9	27	27.5	29.3	trend=	-0.28	IA	ROCKWELL CITY
28.9	29.1	28.3	27.3	29.1	trend=	-0.34	IA	IOWA FALLS
29	29.6	28.6	28.2	29.4	trend=	-0.35	IA	ALBIA 3 NNE
29.6	30.2	28.6	28.8	29.8	trend=	-0.35	IA	LOGAN
28.5	28.9	27.2	26.6	27.9	trend=	-0.37	IA	FORT DODGE 5NNW
29.8	30.3	27.6	29.2	29.9	trend=	-0.43	IA	CLARINDA
36.6	35.4	35.1	35.6	33.8	trend=	1.10	ID	CAMBRIDGE
36.7	34.7	35	34.1	33.3	trend=	0.90	ID	FENN RS
33.8	33.8	32.6	32.5	32.4	trend=	0.89	ID	MALAD CITY
30.4	29.9	29.4	29.4	29.7	trend=	0.83	ID	GRACE
34.1	29.6	30.7	31.3	29.9	trend=	0.80	ID	COEUR D'ALENE
35.5	31.6	32	32.3	31.4	trend=	0.80	ID	PORTHILL
32.6	32	30.5	30	30.6	trend=	0.78	ID	BERN
34	29.7	31	31.4	29.7	trend=	0.77	ID	PRIEST RIVER EXP STN
31.9	31.2	29.7	29.5	30	trend=	0.75	ID	ASHTON 1N
29.6	29.7	28.8	28	28.4	trend=	0.75	ID	LIFTON PUMPING STN
33.2	31.3	31.2	33.2	30	trend=	0.75	ID	NEW MEADOWS RS
32.9	30.2	30.3	30.8	29.7	trend=	0.74	ID	SANDPOINT EXP STN
33.8	32.9	32.2	32.3	31.5	trend=	0.71	ID	ABERDEEN EXP STN
36.3	35.2	35	35.8	33.4	trend=	0.71	ID	ARROWROCK DAM
35.2	34.7	33.9	33.7	32.8	trend=	0.71	ID	HAZELTON
35.9	34.6	34.7	35.4	33	trend=	0.66	ID	DWORSHAK FISH HATCHERY
37.2	36.2	35.8	36.2	34.2	trend=	0.66	ID	NAMPA SUGAR FACTORY
39.5	38.3	38.3	38.3	37	trend=	0.63	ID	GLENNS FERRY
35	34.2	33.5	33.4	32	trend=	0.62	ID	JEROME
36.4	34.9	34.7	35.3	33.4	trend=	0.58	ID	PAYETTE
31.9	32	31.1	31.2	30.4	trend=	0.56	ID	OAKLEY
29	28.1	27.9	28.4	27.3	trend=	0.54	ID	KETCHUM RS
32.8	29.2	31.3	32.2	30.1	trend=	0.52	ID	MOSCOW U OF I
33.3	32.6	32.2	32.5	30.9	trend=	0.51	ID	MAY 2SSE

34.4	32.8	32.9	33.4	31.2 trend=	0.50 ID	SALMON-KSRA
32.7	32.2	31.6	31.7	31.3 trend=	0.49 ID	HOLLISTER
36.5	34.1	35.2	36.2	34.1 trend=	0.49 ID	LEWISTON AP
32	31.6	29.9	30.4	29.8 trend=	0.49 ID	MACKAY LOST RIVER RS
35.3	32.2	32.3	33.2	30.9 trend=	0.33 ID	KELLOGG
30.1	30.4	30.7	30.5	32.3 trend=	0.14 IL	CHARLESTON
30	30.6	31.1	29.5	32.4 trend=	0.09 IL	PANA 3E
27.6	28.2	28.4	27.6	28.9 trend=	0.06 IL	MT CARROLL
28.6	29.3	29.8	28.4	30.8 trend=	0.02 IL	PERRY 6 NW
27.3	27.9	28.5	27.4	29.3 trend=	-0.03 IL	DIXON 1 NW
27.4	28.3	28.8	27.4	29.4 trend=	-0.04 IL	MORRISON
29.3	31.2	30.8	29.6	31.5 trend=	-0.05 IL	HILLSBORO
29.4	30.2	30.5	29.2	31.6 trend=	-0.05 IL	JACKSONVILLE 2E
28.2	30	30	28.5	31.2 trend=	-0.05 IL	URBANA
30.6	31.8	32	31.1	33 trend=	-0.07 IL	MCLEANSBORO
27.2	27.7	28.1	26.9	28.9 trend=	-0.07 IL	MONMOUTH
27.6	26.7	28.5	27.7	30.1 trend=	-0.08 IL	AURORA
28.8	29.8	30.3	29.1	31.6 trend=	-0.10 IL	PALESTINE
29.2	30.5	30.5	29	31.1 trend=	-0.11 IL	CARLINVILLE
29.1	29.5	29.6	28.8	30.9 trend=	-0.11 IL	PARIS WTR WKS
28.1	28.6	29.3	28.1	30.1 trend=	-0.12 IL	MARENGO
29.9	31.8	31.8	30	32.4 trend=	-0.12 IL	SPARTA 1 W
27.7	28.2	28	27	28.9 trend=	-0.14 IL	GALVA
29.9	30.5	30.9	29.6	32 trend=	-0.14 IL	OLNEY 2S
28.2	29	29.3	27.9	30.5 trend=	-0.15 IL	WINDSOR
28.7	30.4	29.7	28.2	30.5 trend=	-0.16 IL	ANNA 2 NNE
30.5	32.4	31.2	31	32.9 trend=	-0.16 IL	DU QUOIN 4 SE
28.9	29.9	30	28.6	30.9 trend=	-0.16 IL	WHITE HALL 1 E
27.5	28.6	28.9	28.4	30 trend=	-0.17 IL	PONTIAC
30.7	32.8	32	30.9	32.6 trend=	-0.18 IL	HARRISBURG
27.5	29.1	28.5	27.7	29.8 trend=	-0.18 IL	HOOPESTON 1 NE
29	29.3	29.7	29.1	31.4 trend=	-0.18 IL	MINONK
28.2	29.3	29.4	27.5	29.6 trend=	-0.21 IL	ALEDO
27.9	28.9	28.9	28.1	30.7 trend=	-0.21 IL	DANVILLE
28.2	28.4	29.7	28.2	30.5 trend=	-0.22 IL	LA HARPE
28.4	28.9	29.1	28.3	30.2 trend=	-0.22 IL	LINCOLN
27.8	29.5	29.2	28.3	30 trend=	-0.23 IL	DECATUR WTP
26.9	27.6	27.9	26.9	29.4 trend=	-0.23 IL	OTTAWA 5SW
27.8	28.2	28.2	27.7	29.5 trend=	-0.24 IL	WALNUT
28.1	29	29.4	28.1	30.1 trend=	-0.31 IL	RUSHVILLE
30.2	30.9	31.3	30.1	33.4 trend=	-0.33 IL	MT VERNON 3 NE
27.6	28.5	28.6	27.1	30 trend=	0.18 IN	WINAMAC 2SSE
29.4	30.4	29.9	29.7	31.4 trend=	0.12 IN	SEYMOUR 2 N
29.7	31	30.4	30.2	31.4 trend=	0.05 IN	CHARLESTOWN 5 NNW
30.4	31.6	30.9	31	32.8 trend=	0.05 IN	MADISON SEWAGE PLT
30.5	31.5	31	31.3	32.8 trend=	0.05 IN	VEVAY
27.9	29.1	28.7	28.1	31 trend=	0.01 IN	ANDERSON SEWAGE PLT

26.9	27.5	27.6	26.9	29.6	trend=	0.01	IN	ANGOLA
27.8	29.6	28.5	28	30.4	trend=	0.01	IN	RUSHVILLE
27.5	29	28.6	27.3	30.3	trend=	0.00	IN	ROCHESTER
30.1	31.1	30.8	30.4	32.4	trend=	-0.02	IN	BROOKVILLE
29.6	30.8	30.4	29.8	31.9	trend=	-0.02	IN	PAOLI
28.3	30.7	30.5	28.9	31.4	trend=	-0.02	IN	ROCKVILLE
27.5	28.4	28.5	27.6	29.5	trend=	-0.03	IN	GOSHEN 3SW
28.3	29.4	29.5	28.1	31.1	trend=	-0.04	IN	WHEATFIELD
31.1	31.7	31.7	30.5	33.2	trend=	-0.05	IN	SHOALS 8 S
27.2	28.3	28.1	27	29.2	trend=	-0.06	IN	WHITESTOWN
26.7	28.5	28	27.4	29.7	trend=	-0.07	IN	CAMBRIDGE CITY 3 N
30.5	31.4	31.5	30.5	33.1	trend=	-0.07	IN	VINCENNES 5 NE
30.7	31	31.3	30.5	32.5	trend=	-0.08	IN	PRINCETON 1 W
28.5	29.8	29.1	28.9	31.1	trend=	-0.09	IN	BLOOMINGTON IN UNIV
28.5	29.7	29.5	28.2	30.7	trend=	-0.09	IN	MARION 2 N
29.9	31.2	30.8	30.5	32.3	trend=	-0.09	IN	SCOTTSBURG
28.3	28.7	28.5	27.6	29.7	trend=	-0.10	IN	BERNE WWTP
28.3	29.8	29.7	28.2	31.3	trend=	-0.10	IN	DELPHI 2 N
28.8	30.1	29.7	29.4	31.3	trend=	-0.10	IN	SALEM
28.3	30.4	29.1	28.2	31	trend=	-0.11	IN	GREENFIELD
28	29	28.9	28	30.6	trend=	-0.11	IN	HUNTINGTON
27.6	28.9	29	27.8	30.9	trend=	-0.11	IN	RENSSELAER
26.8	27.9	27.7	27	28.8	trend=	-0.12	IN	HOBART 2 WNW
30.2	31.9	31.7	30.9	33.1	trend=	-0.13	IN	MT VERNON
29.4	30.7	30.8	30.3	32.2	trend=	-0.13	IN	WASHINGTON 1 W
26.7	27.9	27.8	27	29.2	trend=	-0.14	IN	LAPORTE
28.7	29.6	29.2	28.7	30.8	trend=	-0.15	IN	OOLITIC PURDUE EX FRM
29.1	30.9	30.2	29.9	31.3	trend=	-0.22	IN	COLUMBUS
27.7	29	28.6	27.4	30.1	trend=	-0.24	IN	CRAWFORDSVILLE 6 SE
27.7	29.1	28.6	27.9	29.9	trend=	-0.31	IN	GREENCASTLE 1 W
33.9	38.2	35.2	33.1	34.8	trend=	0.29	KS	JETMORE 8NNW
33.7	36.3	33.5	34.2	34.1	trend=	0.15	KS	LAKIN
32.7	34.7	33.3	33.4	33.9	trend=	0.12	KS	WAKEENEY
33.1	35.1	32	32.8	33.5	trend=	0.08	KS	OBERLIN
32.2	34.2	32.9	33	32.9	trend=	0.05	KS	LARNED
33.5	35	33.3	34	34	trend=	0.05	KS	SAINT FRANCIS
33.9	38	33.8	34.1	35.2	trend=	0.01	KS	ASHLAND
32.3	34.9	32.7	32.8	33.1	trend=	0.01	KS	SCOTT CITY
32.4	37	32.6	34	33.5	trend=	0.00	KS	LIBERAL
32.6	35.4	33.2	33.2	33.5	trend=	-0.02	KS	HAYS 1 S
30.8	31.6	31	30.9	31.4	trend=	-0.03	KS	ATCHISON
32.9	33.8	34.8	34.3	34.1	trend=	-0.03	KS	MINNEAPOLIS
32.8	33.8	32.3	33	33.1	trend=	-0.04	KS	SMITH CTR
30.4	32.3	31.2	30.6	31.7	trend=	-0.10	KS	OLATHE 3E
32.7	34.5	31.7	33.1	33.3	trend=	-0.11	KS	NORTON 9SSE
33.8	34.1	33.8	34.1	34.3	trend=	-0.12	KS	ELLSWORTH
30.6	35.7	32.8	31.9	33	trend=	-0.13	KS	COLUMBUS

31.1	32.3	31.8	31	32.2 trend=	-0.16 KS	LEAVENWORTH
30.9	31.4	31.9	30.9	32.4 trend=	-0.17 KS	MANHATTAN
31.8	37.1	33.8	33.8	33.3 trend=	-0.18 KS	SEDAN
31.4	36.1	33.2	32.5	32.9 trend=	-0.19 KS	INDEPENDENCE
30.9	35.2	32	31.6	32.1 trend=	-0.21 KS	COLDWATER
32.8	35	34	33.2	33 trend=	-0.21 KS	MCPHERSON
30.6	31.7	31.4	31.4	32.4 trend=	-0.22 KS	LAWRENCE
32.8	36.9	33.5	33.5	33.9 trend=	-0.22 KS	MEDICINE LODGE
32	35.1	33.8	33.5	33.6 trend=	-0.24 KS	EL DORADO
30.4	34.2	32.6	31.1	32.9 trend=	-0.25 KS	FT SCOTT
30	31.8	31.9	30.6	31.5 trend=	-0.27 KS	COUNCIL GROVE LAKE
32.3	36.3	33.2	33.3	33.3 trend=	-0.34 KS	ANTHONY
30.4	30.9	30.8	30.9	31 trend=	-0.35 KS	HORTON
30.3	32.3	31.6	30.8	32 trend=	-0.36 KS	OTTAWA
29.7	30.6	31.1	31.4	32.4 trend=	0.28 KY	ASHLAND
29.4	30.5	30.1	30.5	31.4 trend=	0.27 KY	WILLIAMSTOWN 3 W
29.8	31.1	30.4	31.1	32.1 trend=	0.09 KY	FRANKFORT DOWNTOWN
30.5	31.7	31	31.4	32.5 trend=	0.09 KY	SHELBYVILLE 1 E
28.3	29.5	28.5	29.2	30.4 trend=	0.08 KY	WILLIAMSBURG
29.5	31.5	31	29.8	32.2 trend=	0.07 KY	HENDERSON 8 SSW
30.9	32.2	31.1	31.4	33 trend=	0.06 KY	BARBOURVILLE
29.5	30	29.4	30	30.9 trend=	0.05 KY	FARMERS 2 S
30.7	32.2	31.4	31.6	32.7 trend=	0.04 KY	GREENSBURG
29.3	30.4	29.9	30.6	31.4 trend=	-0.02 KY	BEREA COLLEGE
31.6	33.3	31.7	31.8	33.6 trend=	-0.05 KY	BOWLING GREEN RGNL AP
29.8	32	30.2	30.3	32 trend=	-0.15 KY	HOPKINSVILLE
29	30.6	29	29.3	31.3 trend=	-0.19 KY	LEITCHFIELD 2 N
32.3	32.8	35	34	34.6 trend=	0.35 LA	NEW ORLEANS AUDUBON
32.5	34.1	35.3	33.7	34.1 trend=	0.17 LA	BATON ROUGE METRO AP
32.5	33.8	35	32.7	33.1 trend=	0.16 LA	THIBODAUX 3 ESE
32.5	33.2	35	33.4	34.1 trend=	0.15 LA	HOUMA
33	33.4	35.7	33.8	33 trend=	0.13 LA	LAFAYETTE FCWOS
31.9	32.8	35.3	33.8	34 trend=	0.12 LA	COVINGTON 4 NNW
32.6	33.1	34.9	33	33.2 trend=	0.11 LA	DONALDSONVILLE 4 SW
32.1	33.3	34.1	32.4	33.3 trend=	0.09 LA	GRAND COTEAU
32.5	35.6	34.6	32.8	34.6 trend=	0.07 LA	CALHOUN RSCH STN
33	35.3	35.3	33.1	34.9 trend=	0.06 LA	WINNSBORO 5 SSE
33.1	36.6	34	32.5	34.4 trend=	0.05 LA	PLAIN DEALING 4 W
31.3	31.5	33.8	31.9	32.8 trend=	0.04 LA	FRANKLIN 3 NW
31.8	32.2	34.5	31.7	32.9 trend=	0.02 LA	JENNINGS
32.1	34	35.6	33	34.7 trend=	0.01 LA	ALEXANDRIA
31.7	34.2	33.7	31.8	33.5 trend=	-0.02 LA	BASTROP
32.9	35.6	35.2	33.7	35.1 trend=	-0.05 LA	ST JOSEPH 3 N
32.5	33.7	35	33.5	34.5 trend=	-0.07 LA	AMITE
32.5	33.7	35	32.2	33.9 trend=	-0.11 LA	BUNKIE
27.3	30.3	29.4	29.1	29.8 trend=	0.59 MA	NEW BEDFORD
25.5	30.6	28.6	29.8	30 trend=	0.33 MA	BEDFORD

26.7	29.8	29.2	29.4	30.1 trend=	0.30 MA AMHERST
26.4	29.8	29	29.2	29.6 trend=	0.29 MA GREAT BARRINGTON 5 SW
26	31.1	29	31.5	30.5 trend=	0.26 MA LAWRENCE
24.1	28.8	27.1	27.2	27.7 trend=	0.25 MA PROVINCETOWN
25.7	30.4	29.2	29.3	30.3 trend=	0.25 MA WALPOLE 2
26.4	30.3	29.3	29	30.1 trend=	0.22 MA TAUNTON
25	30.3	28.3	28.8	29.8 trend=	0.21 MA BLUE HILL
26	31.1	29.3	30.2	30.5 trend=	0.20 MA READING
26.3	31	29.5	29.9	30.4 trend=	0.19 MA PLYMOUTH-KINGSTON
26.5	31.4	29.8	30.4	30.8 trend=	0.13 MA WEST MEDWAY
30.8	31.1	30.7	31.6	31.4 trend=	0.50 MD CHESTERTOWN
31.1	31.4	31.3	31.6	31.6 trend=	0.46 MD MILLINGTON 1 SE
31.8	31.5	32.1	33.2	32.6 trend=	0.45 MD WOODSTOCK
31.4	31	31.9	32.8	32.6 trend=	0.44 MD WESTMINSTER POL BRKS
31.2	31.2	31.4	31.7	32 trend=	0.42 MD ROYAL OAK 2 SSW
31	31.2	31.4	31	31.8 trend=	0.42 MD SALISBURY
31.9	31.3	32	32.9	31.7 trend=	0.40 MD LAUREL 3 W
31.8	31.4	31.8	32.8	32 trend=	0.38 MD GLENN DALE BELL STN
31.5	32	31.6	33.6	31.7 trend=	0.37 MD CUMBERLAND 2
33.7	33.4	33.2	33.8	35 trend=	0.37 MD MD SCI CTR BALTIMORE
31.3	31.3	31.7	32.7	31.9 trend=	0.36 MD BELTSVILLE
31.4	31.8	32.1	32.1	32.3 trend=	0.30 MD CAMBRIDGE WATER TRMT P
31.4	31.7	32	33	32.3 trend=	0.30 MD DENTON 2 E
30.9	30.7	31	31.9	31.3 trend=	0.29 MD OWINGS FERRY LANDING
32	32.4	32.4	32.3	33.3 trend=	0.27 MD PRINCESS ANNE
27.6	27.1	27.3	29	29.3 trend=	0.14 MD OAKLAND 1 SE
23.6	27.7	27.4	27.6	26.8 trend=	0.28 ME MILLINOCKET
24.8	28.3	27	26.9	27.6 trend=	0.28 ME WOODLAND
24.1	27.1	27.7	27.4	26.4 trend=	0.21 ME HOULTON 5N
24.2	28.4	27.4	28.5	28 trend=	0.20 ME LEWISTON
22.6	25.5	26.2	26.5	25.6 trend=	0.15 ME BRASSUA DAM
24.1	26.8	27.5	27.1	26.5 trend=	0.15 ME PRESQUE ISLE
23.4	28.5	26.2	28	27 trend=	0.11 ME PORTLAND JETPORT
21	25	24.3	24.2	23.8 trend=	0.10 ME EASTPORT
23.1	27.7	26.7	27.8	26.7 trend=	0.06 ME FARMINGTON
24.1	28.2	26.9	28.7	27.3 trend=	0.04 ME GARDINER
22.9	27.2	26	26.5	26 trend=	0.03 ME ACADIA NP
24.2	28.5	28.1	28.4	27.8 trend=	0.00 ME CORINNA
27.1	29.3	28.3	27.1	28.5 trend=	0.32 MI MT PLEASANT UNIV
24.2	23.3	23.7	24.3	24.6 trend=	0.25 MI NEWBERRY 3S
26.5	27.3	27	26.6	28.6 trend=	0.22 MI HART 3 WSW
26	25.9	25.4	25.6	25.8 trend=	0.21 MI CHATHAM EXP FARM 2
27.6	28.4	28.2	28	29.7 trend=	0.19 MI HILLSDALE
27	28.3	26.9	27.3	28.5 trend=	0.18 MI MT CLEMENS ANG BASE
27	27.1	26.5	26.6	27.5 trend=	0.17 MI IRON MT KINGSFORD WWTP
25.6	24.8	24	24.5	25.2 trend=	0.17 MI IRONWOOD
26.7	27.6	27.2	26.6	28.8 trend=	0.14 MI BIG RAPIDS WTR WKS

23.7	23.9	23.4	22.9	24.2	trend=	0.13	MI	FAYETTE 4 SW
24.5	26.2	25.7	25.6	26.8	trend=	0.11	MI	CHEBOYGAN
27.5	28.6	27.6	27.9	29.1	trend=	0.10	MI	ANN ARBOR U OF
27.5	28.8	27.6	27	28.8	trend=	0.09	MI	OWOSSO WWTP
24.4	25.3	24.7	24.6	26.4	trend=	0.09	MI	SOUTH HAVEN
27.6	28.3	28.1	28.2	30.1	trend=	0.07	MI	ADRIAN 2 NNE
26.1	26.7	26.1	25.9	26.6	trend=	0.07	MI	EAST TAWAS
25.7	25.4	24.5	24.4	25.4	trend=	0.06	MI	STAMBAUGH 2SSE
24.2	23.5	22.8	23.9	24.4	trend=	0.05	MI	CHAMPION VAN RIPER PK
27.6	28.7	28.4	27.9	29.9	trend=	0.05	MI	KALAMAZOO STATE HOSP
27	27.9	27.6	27	29	trend=	0.04	MI	ALLEGAN 5NE
26.6	27.6	27.5	27	29	trend=	0.01	MI	COLDWATER ST SCHOOL
26.9	28.4	27.6	27.2	29	trend=	-0.01	MI	ALMA
25.1	26.8	25.4	25.7	27	trend=	-0.05	MI	MIDLAND
23	21.9	21.8	22.9	23.8	trend=	-0.18	MI	MUNISING
24.2	21.5	21.7	23.1	21.7	trend=	0.31	MN	TWO HARBORS
26.4	24.8	24.8	25.6	25.2	trend=	0.24	MN	CLOQUET
29.4	26	26.3	27.8	26.9	trend=	0.24	MN	GRAND RPDS FOREST LAB
29	25.9	25.1	26.7	25.8	trend=	0.24	MN	ROSEAU
28.4	25.6	25.1	26.4	26.3	trend=	0.22	MN	LEECH LAKE
27.3	24.9	24.3	26.4	25.5	trend=	0.17	MN	EVELETH WWTP
28	25.8	25.7	26.3	25.7	trend=	0.15	MN	SANDY LAKE DAM LIBBY
28.8	26.1	25.6	27.4	26.3	trend=	0.10	MN	ITASCA UNIV OF MINN
30.9	29.5	29.8	28.7	29.4	trend=	0.08	MN	CHASKA
30	29.6	29.5	28.1	29.4	trend=	0.08	MN	MINNEAPOLIS/ST PAUL AP
29.2	26.8	26.2	27.2	27.5	trend=	0.07	MN	PINE RIVER DAM
30.4	27	25.5	29	26.9	trend=	0.06	MN	BAUDETTE
28	25.8	25.5	26.2	25.8	trend=	0.05	MN	WALKER AH GWAH CHING
27.5	24.9	24.2	25.8	25.3	trend=	0.04	MN	MARCELL 5NE
29.1	27.1	26.8	27.7	26.3	trend=	0.03	MN	PARK RAPIDS 2 S
28.8	26.6	25.6	26.8	25.8	trend=	0.01	MN	FOSSTON 1 E
28.5	29.2	27	28.1	27.3	trend=	-0.02	MN	MORRIS WC EXP STN
30.8	29.8	28.9	28.5	28.9	trend=	-0.07	MN	MILAN 1 NW
30.9	28.5	27	28.3	27.7	trend=	-0.15	MN	ARGYLE
28.2	28.6	27.3	27.4	28.2	trend=	-0.19	MN	FAIRMONT
28.9	28.3	28	26.9	28.3	trend=	-0.20	MN	FARMINGTON 3 NW
28.5	27.5	27.3	27.4	28.2	trend=	-0.21	MN	MORA
29.9	28	25.5	27	26.5	trend=	-0.24	MN	ADA
26.8	26.7	26.9	25.7	27.3	trend=	-0.25	MN	GRAND MEADOW
28.6	26.9	25.9	27.3	26.1	trend=	-0.26	MN	DETROIT LAKES 1 NNE
28.9	28.8	27.7	27.7	28.5	trend=	-0.27	MN	WINNEBAGO
27.5	27.9	28.2	27	28.5	trend=	-0.29	MN	ZUMBROTA
28	28	27.1	26.8	27.7	trend=	-0.30	MN	ALBERT LEA 3 SE
28.6	28.3	28	27.1	28.4	trend=	-0.32	MN	ST PETER
27.2	27.4	26.2	26.1	27.1	trend=	-0.35	MN	NEW ULM 2 SE
28.7	28.6	27.8	27.3	27.8	trend=	-0.35	MN	OLIVIA 3E
28.8	28.7	27.6	27.7	28	trend=	-0.36	MN	MONTEVIDEO 1 SW

28.7	28.9	27.6	27.4	28.3	trend=	-0.44	MN	PIPESTONE
29.6	30.7	31.5	29.2	31	trend=	0.01	MO	BRUNSWICK
31	32.8	31.6	30.2	32.2	trend=	0.01	MO	FARMINGTON
29.9	32.2	32.3	29.3	31.7	trend=	0.01	MO	WARRENTON 1 N
28.6	30.4	31.6	28.1	30.3	trend=	-0.05	MO	BOWLING GREEN 1 E
30.5	32.8	32.6	29.9	32.3	trend=	-0.13	MO	JEFFERSON CITY WTP
30.1	34.7	33.2	30.7	32.1	trend=	-0.14	MO	LAMAR
30.5	35.1	33.1	30.3	32	trend=	-0.14	MO	LEBANON 2W
29.1	34.1	30.7	29.1	31.5	trend=	-0.14	MO	MTN GROVE 2 N
30.7	33.4	32.6	30.5	32.5	trend=	-0.16	MO	MARBLE HILL
30.4	34.6	31.5	29.6	32.3	trend=	-0.17	MO	ROLLA UNI OF MISSOURI
31.4	34	31.8	31.2	33.4	trend=	-0.19	MO	CARUTHERSVILLE
28.7	30.4	29.9	28.9	30.4	trend=	-0.19	MO	LEES SUMMIT REED WR
30.1	31.8	31.9	29.5	31.6	trend=	-0.19	MO	MEXICO
28.7	29.5	31.4	28.4	30.3	trend=	-0.19	MO	STEFFENVILLE
29.2	29.5	29.7	28.9	29.8	trend=	-0.20	MO	SPICKARD 7 W
30.5	35.1	32.8	30.5	33	trend=	-0.21	MO	LOCKWOOD
30.9	36	33.6	32.3	32.6	trend=	-0.22	MO	NEOSHO
30.4	33	32.5	30.3	32.3	trend=	-0.26	MO	CLINTON
31.7	34.5	32.6	31.4	33.8	trend=	-0.28	MO	DONIPHAN
30.3	33.2	32.4	30.2	32.2	trend=	-0.28	MO	TRUMAN DAM & RSVR
29.8	29.6	29.4	29.1	29.9	trend=	-0.29	MO	CONCEPTION
30.5	32.2	32.2	30	32	trend=	-0.29	MO	SWEET SPRINGS
28.8	29.2	29.3	28.6	29.7	trend=	-0.30	MO	UNIONVILLE
30	32.7	32.7	30.1	32.1	trend=	-0.31	MO	APPLETON CITY
30.1	31.8	31.8	29.9	31.5	trend=	-0.32	MO	LEXINGTON 3E
29.7	31.5	31.8	29.3	31.5	trend=	-0.33	MO	MOBERLY
31.7	32.7	33.7	32.6	33.3	trend=	0.27	MS	POPLARVILLE EXP STN
32.2	34.2	33.8	33.1	33.6	trend=	0.19	MS	LOUISVILLE
32.9	34	35.2	34.3	35.1	trend=	0.16	MS	LAUREL
32.5	34.3	34.5	34.1	34.4	trend=	0.10	MS	CRYSTAL SPGS EXP STN
32.3	34.5	33.7	33.7	34.3	trend=	0.10	MS	STATE UNIV
31.4	33.8	33.2	32.4	33.5	trend=	0.09	MS	KOSCIUSKO
33.1	35.3	34.3	34.7	35.2	trend=	0.05	MS	COLUMBUS
30.8	34	31.8	31.8	33	trend=	0.05	MS	CORINTH 7 SW
31.7	35.1	32.8	32.1	33.9	trend=	0.05	MS	HOLLY SPRINGS 4 N
31	31.8	33.6	32.5	33.2	trend=	0.05	MS	PASCAGOULA 3 NE
30.1	32.7	31.7	31.6	31.9	trend=	0.04	MS	BOONEVILLE
31.2	32.8	33.8	32.8	33.2	trend=	0.03	MS	FOREST
34.5	35.9	36.8	35.8	36.9	trend=	0.03	MS	MONTICELLO
32.5	33.9	35.2	33.7	34.5	trend=	0.02	MS	COLUMBIA
31.7	33.9	34.2	33.6	34.4	trend=	0.02	MS	PORT GIBSON 1 NE
33	34.4	35.5	33.6	34.5	trend=	0.02	MS	WOODVILLE 4 ESE
31	34	34.2	33	33.4	trend=	-0.01	MS	CANTON 4N
31.3	33.4	33.2	32.3	33.3	trend=	-0.02	MS	NATCHEZ
31.4	32.7	33.6	32.3	33.4	trend=	-0.04	MS	BROOKHAVEN CITY
31.9	33.8	33.4	31.3	34	trend=	-0.04	MS	GREENVILLE

31.9	33.4	34.6	33.6	34.1 trend=	-0.04 MS	HATTIESBURG 5SW
31.6	34	33.2	32	33.6 trend=	-0.05 MS	WATER VALLEY
33.5	35.7	35.5	35	35.9 trend=	-0.06 MS	YAZOO CITY 5 NNE
31.6	34.4	31.8	33.2	34.1 trend=	-0.07 MS	ABERDEEN
32.3	34.8	33.6	33.4	34.4 trend=	-0.07 MS	UNIVERSITY
30.7	31.4	33	31.8	32.3 trend=	-0.08 MS	WAVELAND
32.7	34.1	34.6	34.2	34.6 trend=	-0.08 MS	WAYNESBORO 2 W
30.9	33.7	32.4	31.8	33.2 trend=	-0.10 MS	PONTOTOC EXP STN
31.7	34.3	32.9	33.1	34.3 trend=	-0.14 MS	BATESVILLE 2 SW
31.7	33.4	32.8	31.7	33.9 trend=	-0.18 MS	MOORHEAD
32.9	34.9	34.1	32.8	34.7 trend=	-0.22 MS	CLARKSDALE
31	33.1	32.2	31.7	33.3 trend=	-0.35 MS	HERNANDO
32.4	30.8	31.1	31.2	29.6 trend=	0.86 MT	HAMILTON
30.8	28.7	28.5	29.8	27.8 trend=	0.82 MT	DILLON WMCE
35.7	32.7	33.4	33.6	31.8 trend=	0.70 MT	LIBBY 1 NE RS
35	33.1	31	32.7	30.7 trend=	0.68 MT	HUNTLEY EXP STN
32.9	30.8	30.6	32	29.2 trend=	0.68 MT	SAINT IGNATIUS
30.9	28	28.9	30.2	24.8 trend=	0.65 MT	CUT BANK AP
30.4	28.5	28	29.1	27 trend=	0.65 MT	PHILIPSBURG RS
29.3	28.5	26.5	28	25.8 trend=	0.64 MT	ANACONDA
32.7	29.4	30.2	30.4	28.1 trend=	0.64 MT	FORTINE 1 N
31.9	29.7	28.6	29.3	27.6 trend=	0.64 MT	MOCCASIN EXP STN
30.5	29.5	28.1	27.7	28 trend=	0.64 MT	WEST YELLOWSTONE
32.2	30.2	29	30.5	29.1 trend=	0.63 MT	ENNIS
29.6	28.1	28.3	29	28.4 trend=	0.63 MT	VIRGINIA CITY
34.4	32.4	32.2	32.5	30.4 trend=	0.61 MT	SAINT REGIS 1 NE
35.1	32.9	31.1	33.3	28.8 trend=	0.60 MT	GLASGOW INTL AP
33	31.6	30.6	31.9	29.2 trend=	0.59 MT	CASCADE 5 S
33	29.6	30.6	31.1	27.7 trend=	0.59 MT	KALISPELL GLACIER AP
32.9	30.9	29	31.7	29.4 trend=	0.57 MT	FLATWILLOW 4 ENE
34.7	32.2	31.6	33.1	29.7 trend=	0.54 MT	HELENA AP ASOS
30.1	29.1	27.5	28.8	27.4 trend=	0.53 MT	BOZEMAN MONTANA ST U
30.2	28.5	26.2	28.1	26.7 trend=	0.53 MT	RED LODGE
34.1	32.3	30	31.9	30.1 trend=	0.49 MT	HYSHAM 25 SSE
30.2	28.6	27.2	28.4	27.1 trend=	0.49 MT	WHITE SULPHUR SPRNGS 2
32.1	30.4	29.4	30	29.8 trend=	0.47 MT	NORRIS MADISON PH
29	27.4	26.7	28	25.2 trend=	0.45 MT	AUGUSTA
27.6	27.6	26.1	26.6	26.5 trend=	0.44 MT	HEBGEN DAM
33.3	31.4	30.3	32.4	28.4 trend=	0.41 MT	CHINOOK
33	31.4	30.1	31	28.6 trend=	0.41 MT	GREAT FALLS AP
33.4	31.4	29.7	31.2	29.3 trend=	0.38 MT	BIG TIMBER
30.2	28.2	27.9	28.8	25.6 trend=	0.38 MT	VALIER
35	32.8	30.9	32.2	29.8 trend=	0.35 MT	MALTA 7 E
29.4	28	27	27.9	27.3 trend=	0.33 MT	LIVINGSTON 12 S
35.5	33.2	31.4	32.4	31.2 trend=	0.33 MT	MILES CITY AP
35.2	32.4	30.6	32.8	30.1 trend=	0.32 MT	JORDAN
33.4	31.1	29.5	31.9	28 trend=	0.31 MT	FORKS 4 NNE

30	29.2	28	29.5	26.7 trend=	0.25 MT	CHOTEAU
33.6	30.4	28.4	30.5	29.4 trend=	0.18 MT	EKALAKA
32.8	29.8	28.7	31.8	27.4 trend=	0.17 MT	MEDICINE LAKE 3 SE
32.5	29.2	29.2	29.5	27.5 trend=	0.16 MT	FT ASSINNIBOINE
35.1	32.1	30.5	32.1	30.6 trend=	0.12 MT	MILDRED 5 N
33.1	30.4	27.7	31.2	29.3 trend=	0.09 MT	GLENDIVE
34.1	30.4	29.3	31.3	29.1 trend=	0.03 MT	PLEVNA
34.6	31	30.1	31.9	30.2 trend=	-0.01 MT	SAVAGE
33.8	31.4	29.9	32.2	28.8 trend=	-0.11 MT	VIDA 6 NE
26.8	27.6	27.2	26.9	27.9 trend=	0.37 NC	TRANSOU
30.4	31.3	31.8	31.1	31.9 trend=	0.33 NC	EDENTON
31.8	32.3	32.5	32.4	33.8 trend=	0.32 NC	MONROE 2 SE
30.3	30.4	31	30.5	31.5 trend=	0.30 NC	CAPE HATTERAS AP
31.2	31.9	31.4	32	32.8 trend=	0.29 NC	LENOIR
31.2	32.6	33	33.4	33.2 trend=	0.28 NC	LOUISBURG
32.9	35.2	34.8	33.8	33.7 trend=	0.27 NC	SMITHFIELD
31.4	32.6	32	32.2	32.8 trend=	0.23 NC	CHAPEL HILL 2 W
31.1	31.8	31.5	31.5	32.3 trend=	0.23 NC	STATESVILLE 2 NNE
32.4	32.8	33.3	32.8	33.9 trend=	0.21 NC	LUMBERTON
30.7	31.7	31.3	30.7	31.8 trend=	0.18 NC	ELIZABETH CITY
24.2	25.8	25.2	25	26.1 trend=	0.18 NC	HIGHLANDS
31	32.2	31.7	31.9	32.4 trend=	0.18 NC	REIDSVILLE 2 NW
32.6	32.8	32.4	31.2	32.5 trend=	0.17 NC	HENDERSON 2 NNW
32.3	33	33.8	32.9	33.7 trend=	0.17 NC	SOUTHPORT 5 N
31.3	31.7	31.5	32.2	32.9 trend=	0.16 NC	MORGANTON
30.7	31.2	31.2	31.4	32 trend=	0.16 NC	MT AIRY 2 W
27.7	28.8	29.5	28.9	29.5 trend=	0.14 NC	MOREHEAD CITY 2 WNW
30.9	32.8	32.8	31.3	32.9 trend=	0.14 NC	TARBORO 1 S
30.6	31.7	32.2	31.6	32.4 trend=	0.11 NC	ALBEMARLE
27.1	28.4	27.5	28.1	29 trend=	0.11 NC	WAYNESVILLE 1 E
31.5	32.3	32.6	32.2	32.9 trend=	0.10 NC	KINSTON 7 SE
31	32	31.5	31.5	32.9 trend=	0.10 NC	SALISBURY
31.2	32.3	32.2	32.1	32.9 trend=	0.08 NC	FAYETTEVILLE PWC
31.5	32.9	32.6	32.2	33.3 trend=	0.08 NC	GOLDSBORO 4 SE
29.4	30.3	30	29.6	30.9 trend=	0.08 NC	MARSHALL
28.8	29.8	29.7	29.3	31 trend=	0.05 NC	HENDERSONVILLE 1 NE
30.8	31.5	30.9	31.1	32.1 trend=	0.02 NC	MARION 2 NW
31	32.8	32.8	32.5	33.5 trend=	-0.01 NC	WILSON 3 SW
31.3	28.6	26.7	29.2	27.5 trend=	0.23 ND	RICHARDTON ABBEY
32.7	29.9	28.2	31	28.2 trend=	0.22 ND	MOTT
32.3	30.8	28.3	30.3	28.3 trend=	0.07 ND	HETTINGER
27.9	25.3	24.2	25.5	24.1 trend=	0.07 ND	LANGDON EXP FARM
31.9	28.9	27.4	29.9	27.8 trend=	0.06 ND	NEW ENGLAND
29.6	26.9	25.5	27.5	26.5 trend=	0.05 ND	BOTTINEAU
32.1	28.9	27.5	29.9	28.2 trend=	0.04 ND	DICKINSON EXP STN
30.2	27.7	26.4	27.8	26.5 trend=	0.03 ND	GRAND FORKS UNIV NWS
31.6	28.1	27.8	28.3	26.9 trend=	0.03 ND	TOWNER 2 NE

29.2	27.5	25.9	26.5	27.2	trend=	0.02	ND	MOFFIT 3 SE
31	27.6	26.9	27.6	26.2	trend=	0.02	ND	WILLOW CITY
29.6	26.7	25.7	27	25.2	trend=	0.00	ND	PEMBINA
33.1	30.8	29.5	31.5	29.1	trend=	-0.02	ND	DUNN CENTER 1E
30.3	27.9	27	28.6	26.8	trend=	-0.02	ND	GRAFTON
31.2	28.8	27	28.3	27.8	trend=	-0.11	ND	MANDAN EXP STN
30.9	27.3	27.2	28.6	25.4	trend=	-0.14	ND	CROSBY
32.4	30.8	28.9	30.2	29.7	trend=	-0.18	ND	FT YATES 4 SW
30.5	28.6	27.6	28.4	27.8	trend=	-0.20	ND	WAHPETON 3 N
30.9	29.3	27.3	28.6	27.6	trend=	-0.21	ND	LISBON
30.7	28.8	27.2	28.4	27.4	trend=	-0.24	ND	HILLSBORO 3 N
29.6	28.4	26.8	26.9	25.9	trend=	-0.28	ND	NAPOLEON
30.8	28.7	27	28	28.2	trend=	-0.32	ND	JAMESTOWN STATE HOSP
30	27.9	26.6	27.6	27	trend=	-0.35	ND	CASSELTON AGRONOMY FM
29.1	28	26.2	27.2	26.2	trend=	-0.55	ND	FULLERTON 1 ESE
34.2	34.3	30.9	33.5	33.1	trend=	0.30	NE	BRIDGEPORT
30.5	32.4	29.2	29.4	30.1	trend=	0.11	NE	ATKINSON 3SW
31.4	31.8	28.4	31.2	30.7	trend=	0.11	NE	KIMBALL 2NE
30.9	33.2	31	31.7	31.5	trend=	0.11	NE	PURDUM
33.3	34.2	30.9	32.5	32.8	trend=	0.07	NE	LODGEPOLE
31.8	32.4	29.6	31.3	31.1	trend=	0.03	NE	ALLIANCE 1WNW
33.1	33.1	29.6	31.5	31.8	trend=	0.02	NE	HAY SPRINGS 12 S
31.5	32.6	32.4	32.1	32	trend=	0.01	NE	RED CLOUD
31.1	30.1	27.8	30.4	29.8	trend=	-0.01	NE	HARRISON
33.1	34.2	30.7	32.5	33	trend=	-0.05	NE	IMPERIAL
32.1	33.7	30.8	31.7	31.5	trend=	-0.06	NE	MERRIMAN
30.9	33.9	31.7	33.5	33.9	trend=	-0.07	NE	MC COOK
30.3	32	30.3	30	31.3	trend=	-0.07	NE	SAINT PAUL 4N
30.4	31.9	29.3	30	30.8	trend=	-0.11	NE	ASHLAND NO 2
29.4	31.1	29.2	29	30.9	trend=	-0.13	NE	MINDEN
29.1	30.8	29.4	28.6	29.8	trend=	-0.13	NE	NORTH LOUP
31.4	32.2	30.4	30.3	31.6	trend=	-0.15	NE	TEKAMAH
30.1	31.1	29.7	28.8	29.7	trend=	-0.16	NE	BROKEN BOW 2 W
29.4	31.2	29.2	29.2	30.7	trend=	-0.18	NE	HOLDREGE
30.2	32	29	29.9	29.8	trend=	-0.18	NE	STAPLETON 5W
31	31.8	30.6	31.1	30.9	trend=	-0.20	NE	FRANKLIN
30.9	31.7	30	30.1	30.6	trend=	-0.21	NE	CRETE
29.2	30.5	28.3	28	29.1	trend=	-0.22	NE	WAKEFIELD
30.3	31.6	29.8	29.2	30.3	trend=	-0.23	NE	LOUP CITY
30.5	30.6	30	30.3	30.9	trend=	-0.24	NE	BEATRICE 1N
31.5	30.4	31.2	30.2	31.3	trend=	-0.24	NE	FAIRBURY 5S
29.7	31.1	29.5	29.5	30.6	trend=	-0.25	NE	GOTHENBURG
31.7	34	32	31.2	33.3	trend=	-0.27	NE	CURTIS 3NNE
30	31.4	29.8	29.6	30.5	trend=	-0.28	NE	FAIRMONT
29.8	29.9	28.6	29	30	trend=	-0.30	NE	DAVID CITY
31.3	32.6	30.9	31.1	32	trend=	-0.30	NE	YORK
32.3	34	31.4	31.7	33.6	trend=	-0.31	NE	BEAVER CITY

29.1	30.6	28.8	28.1	29.2 trend=	-0.31 NE	GENOA 2 W
29.6	31.2	28.6	28.7	29.3 trend=	-0.31 NE	MADISON
29.6	30.6	28.5	29	29.9 trend=	-0.34 NE	WEeping WATER
29.3	30.5	29.2	29	29.9 trend=	-0.36 NE	GENEVA
31.3	31.1	31	30.9	31.2 trend=	-0.38 NE	HEBRON
30.2	30.8	29.6	30.2	30.7 trend=	-0.39 NE	PAWNEE CITY
28.3	30.4	27.5	27.9	28.2 trend=	-0.40 NE	OAKDALE
29.8	31.3	28.6	29	29.4 trend=	-0.41 NE	HARTINGTON
29.6	31.3	30.2	29.6	29.5 trend=	-0.41 NE	HASTINGS 4N
30.5	31.3	30.1	30.4	31.1 trend=	-0.41 NE	SYRACUSE
30.9	32.1	30.2	30.5	31.1 trend=	-0.47 NE	SEWARD
30	30.2	29.3	29.7	30.6 trend=	-0.48 NE	TECUMSEH 1S
30.1	30.6	29.2	29.8	30.2 trend=	-0.54 NE	AUBURN 5 ESE
28.9	30.2	28.3	27.8	28.8 trend=	-0.58 NE	ALBION
25.5	28.9	28.1	29.2	29.8 trend=	0.31 NH	KEENE
24.4	29.1	27.7	29.2	29 trend=	0.30 NH	DURHAM
22.2	24.7	25.5	26	25.5 trend=	0.29 NH	FIRST CONNECTICUT LAKE
24.4	27.6	27.8	28.5	28.2 trend=	0.25 NH	BETHLEHEM 2
25.1	29.6	28.7	29.5	29.8 trend=	0.02 NH	HANOVER
28.9	30.8	29.9	31	31.2 trend=	0.56 NJ	BELVIDERE BRG
31.2	33	32.2	32.9	32.3 trend=	0.56 NJ	HIGHTSTOWN 2 W
29.1	30.6	30.1	30.8	31.1 trend=	0.51 NJ	BOONTON 1 SE
29.7	30.2	30.3	30.8	30.6 trend=	0.51 NJ	MOORESTOWN
30.4	32.6	31	31.9	31.7 trend=	0.43 NJ	NEW BRUNSWICK 3 SE
29.3	30.6	30	29.7	29.8 trend=	0.41 NJ	LONG BRANCH OAKHURST
29.8	32	30.5	31.9	31.5 trend=	0.39 NJ	FLEMINGTON 5 NNW
30.2	32.4	31.5	32.2	32.5 trend=	0.38 NJ	PLAINFIELD
29.8	31.1	30.6	30.6	31 trend=	0.38 NJ	TOMS RIVER
26.7	26.8	27.2	26.2	27.9 trend=	0.29 NJ	ATLANTIC CITY
28.1	29.3	29	29.9	30.1 trend=	0.28 NJ	CHARLOTTEBURG RSVR
30.3	31.9	31.8	31.8	31.9 trend=	0.26 NJ	INDIAN MILLS 2 W
33.2	35.2	38.6	35.2	35 trend=	0.70 NM	LOS LUNAS 3 SSW
32.3	38.2	39.7	36.3	34.5 trend=	0.58 NM	ROSWELL IND AP
34.6	35	37.1	34.5	34.1 trend=	0.56 NM	AZTEC RUINS NM
32.8	35.5	39.7	34.8	33.4 trend=	0.54 NM	SOCORRO
32.6	37.6	36.7	36.1	34 trend=	0.53 NM	TUCUMCARI 4 NE
22.9	23.9	26.8	23.3	23.1 trend=	0.49 NM	RED RIVER
32.1	34.9	38	34.6	33.3 trend=	0.49 NM	TULAROSA
33.5	36.1	39.3	35.8	35 trend=	0.48 NM	GAGE
24	26.7	30.3	26.6	25.6 trend=	0.47 NM	MTN PARK
33	36.7	39.3	36.3	35.3 trend=	0.45 NM	JORNADA EXP RANGE
30.3	33.3	36.3	32.7	31.3 trend=	0.44 NM	CARRIZOZO 1SW
28.5	29.2	33.3	31.2	30.4 trend=	0.44 NM	LUNA RS
32.3	36.7	36.6	35	33.2 trend=	0.42 NM	BELL RANCH
31	32.5	36.7	33.7	32 trend=	0.42 NM	FT BAYARD
33.7	36.3	39.3	36.1	35.1 trend=	0.41 NM	ELEPHANT BUTTE DAM
32.7	37.9	37.2	35.6	33.8 trend=	0.41 NM	SAN JON

31	36	35.8	33.8	32 trend=	0.40 NM FT SUMNER
31.1	30.3	33.3	30.8	30.3 trend=	0.39 NM DULCE
28.8	30.1	33.7	30.1	29.4 trend=	0.39 NM MOUNTAINAIR
31.7	32.8	34.2	32.5	32.2 trend=	0.38 NM SPRINGER
28.5	30.1	32.3	29.7	28.6 trend=	0.36 NM LAS VEGAS WWTP
34	37	39.7	36.2	35.3 trend=	0.36 NM STATE UNIV
31.4	34.8	35.8	33.6	32.2 trend=	0.35 NM SANTA ROSA
30.9	33.5	32.8	32.3	30.2 trend=	0.33 NM CLAYTON MUNI ARPK AP
33.3	38	39.3	36.3	35 trend=	0.32 NM CARLSBAD
27.6	26.2	30.1	26	26.6 trend=	0.30 NM CHAMA
27.1	28.8	31	30.2	28.8 trend=	0.28 NM CIMARRON 4 SW
33.8	36.6	39.5	36.2	34.7 trend=	0.23 NM OROGRANDE
29.7	31.8	34.2	31.6	31.1 trend=	0.13 NM JEMEZ SPRINGS
36.8	36.9	35.9	37	34.7 trend=	0.90 NV FALLON EXP STN
35.8	35.9	35	35.5	33.9 trend=	0.72 NV ELKO RGNL AP
36.5	35.1	35.7	36.5	33.8 trend=	0.72 NV LOVELOCK
36.5	35.8	36.5	37	33.7 trend=	0.70 NV RENO AP
37.7	36.9	36.7	37.5	34.6 trend=	0.70 NV WINNEMUCCA AP
37.6	37.5	37.9	38.2	35.5 trend=	0.60 NV MINA
32.4	32.3	33.4	32.9	32.5 trend=	0.56 NV MCGILL
38.6	38.8	38.5	39.1	37.3 trend=	0.53 NV BATTLE MOUNTAIN 4SE
33.3	33.4	32.3	32.4	31.6 trend=	0.53 NV WELLS
35.6	34.8	34.7	35.5	33.3 trend=	0.49 NV GOLCONDA
30.5	30.3	31	31	30 trend=	0.40 NV AUSTIN #2
37.8	36.7	39.7	40.1	36.5 trend=	0.34 NV SEARCHLIGHT
39.2	38.5	41	42.7	38.8 trend=	0.31 NV BOULDER CITY
26.2	29.3	28.5	28.4	29.7 trend=	0.54 NY MOHONK LAKE
21	23.6	24.2	24.9	25.3 trend=	0.49 NY LAKE PLACID 2 S
28.4	30.9	30.3	30.9	31.2 trend=	0.44 NY GLENHAM
26	28.6	28.2	27.2	28.5 trend=	0.42 NY BRIDGEHAMPTON
27.1	30.4	29.2	30.1	30.7 trend=	0.41 NY PORT JERVIS
27.6	31.2	28.7	29.6	29.5 trend=	0.37 NY ELMIRA
24.5	26.1	26.9	27.2	27 trend=	0.35 NY MORRISVILLE 6 SW
25.9	28.3	27.4	28.2	28.8 trend=	0.34 NY ITHACA CORNELL UNIV
28.4	30.7	30.2	30.7	30.4 trend=	0.34 NY NY CITY CNTRL PARK
28	30.5	29.6	30.3	30.6 trend=	0.34 NY WEST POINT
26.9	30.8	29.9	30.7	30 trend=	0.32 NY ALBANY INTL AP
26.2	28.8	28	28.9	29 trend=	0.32 NY CORTLAND
27.5	29.7	29.3	28.9	29.9 trend=	0.32 NY SETAUKET STRONG
27.5	29.4	28.2	29	29.7 trend=	0.30 NY ADDISON
25.8	28.2	27.1	27.8	29 trend=	0.29 NY ALFRED
28.2	31.3	30.8	31.6	31.7 trend=	0.29 NY TROY L&D
25	27.4	26.8	27.4	27.3 trend=	0.27 NY BINGHAMTON GREATER AP
24	26.5	26.1	27.4	27.9 trend=	0.27 NY COOPERSTOWN
25.2	27.5	27.8	29.4	29.4 trend=	0.26 NY CHAZY
28.7	30.8	30.2	30.6	31.1 trend=	0.26 NY DOBBS FERRY ARDSLEY
27.4	29.4	29	30.6	29.6 trend=	0.26 NY SYRACUSE WSO AP

24.9	26.7	26.6	26.7	28.8	trend=	0.25	NY	LOCKPORT 3 S
24.7	28.4	26	27.3	27.4	trend=	0.25	NY	NORWICH
26.9	30.2	29.3	29.7	30	trend=	0.25	NY	POUGHKEEPSIE
27.8	30.2	29.5	29.8	30.3	trend=	0.25	NY	YORKTOWN HEIGHTS 1 W
23.9	26.3	26.6	26.9	28	trend=	0.24	NY	LOWVILLE
24.3	26.5	26.9	27.6	28.4	trend=	0.23	NY	LAWRENCEVILLE 3 SW
25.4	27.5	27.8	28.4	28.5	trend=	0.23	NY	UTICA FAA AP
24.7	27	26.7	27.3	28.3	trend=	0.22	NY	GENEVA RSCH FARM
23.1	25.5	26.2	26.7	27.2	trend=	0.22	NY	TUPPER LAKE SUNMOUNT
25.2	27.1	26.8	26.9	28.5	trend=	0.21	NY	BATAVIA
26.7	29.2	27.7	28.5	28.5	trend=	0.21	NY	DEPOSIT
25.8	28.9	28.1	28.5	29.6	trend=	0.21	NY	LITTLE FALLS CITY RSVR
25.2	27.4	27.8	27.9	29	trend=	0.20	NY	CANTON 4 SE
24.6	27	27.1	27.4	28.7	trend=	0.20	NY	WANAKENA RNGR SCHOOL
28.3	30.9	30	30.6	31	trend=	0.19	NY	WALDEN 1 ESE
25.3	27	27	27.1	27.7	trend=	0.18	NY	ALLEGANY SP
27.1	29.4	29.2	29.2	30.8	trend=	0.18	NY	DANVILLE
26.4	29.4	29.1	29.6	30.4	trend=	0.18	NY	LITTLE FALLS MILL ST
24.8	26.7	26.5	27.2	27.9	trend=	0.17	NY	FREDONIA
25.5	27.8	27.5	28	29.2	trend=	0.16	NY	BROCKPORT
22.1	24.3	24.9	25.6	25.9	trend=	0.16	NY	INDIAN LAKE 2SW
26.1	28	27.8	28	29.2	trend=	0.14	NY	ANGELICA
25.6	27.2	27.3	28.4	28.6	trend=	0.13	NY	BUFFALO NIAGARA INTL
24.8	28.1	27.7	28.4	28.9	trend=	0.13	NY	GLOVERSVILLE
24.4	26.4	25.9	26.1	28	trend=	0.13	NY	WATERTOWN
24.7	26.9	25.8	26.8	27	trend=	0.12	NY	MARYLAND 9 SW
24.1	26.5	28.1	27.3	28.8	trend=	0.10	NY	OGDENSBURG 4 NE
25.2	27.9	26.9	27.4	28.6	trend=	0.09	NY	AUBURN
21.7	25.3	24.8	24.9	25.7	trend=	0.09	NY	STILLWATER RSVR
26.1	29.8	29.1	29.5	29.6	trend=	0.08	NY	SARATOGA SPRINGS 4 SW
23.6	26.5	26.8	27	28.2	trend=	0.06	NY	MALONE
25.6	28.2	27.4	28.9	28.9	trend=	0.06	NY	ROCHESTER INTL AP
23.5	25.5	26.3	27.1	27	trend=	0.05	NY	DANNEMORA
25.5	27.4	27.5	27.4	28.7	trend=	0.05	NY	WALES
25.4	27.6	27.4	27.8	29	trend=	-0.01	NY	HEMLOCK
24.2	25.6	26.1	26	27	trend=	-0.04	NY	OSWEGO EAST
27.1	28.7	28.2	28.8	29.7	trend=	0.37	OH	MILLPORT 4 NE
27.3	28.5	28.2	28.2	29.5	trend=	0.32	OH	WARREN 3 S
27	27.8	27.9	28.9	28.8	trend=	0.28	OH	CADIZ
26.2	27	26.6	27.7	28.5	trend=	0.24	OH	HIRAM
26.2	27.6	27.4	27.9	29	trend=	0.18	OH	CHIPPEWA LAKE
28	28.8	28.7	29.2	30	trend=	0.18	OH	WOOSTER EXP STN
29.8	30.6	30.6	31.2	32.2	trend=	0.17	OH	PORTSMOUTH-SCIOTOVILLE
27.8	28.2	29	29.7	29.7	trend=	0.14	OH	COSHOCTON WPC PLT
28.6	30	29.3	30.2	30.2	trend=	0.14	OH	HILLSBORO
26.9	28.8	28.9	28.8	30.1	trend=	0.13	OH	MILLERSBURG
28.2	29.1	28.9	28.7	30.6	trend=	0.12	OH	DEFIANCE

29.6	30.8	30.6	31.8	31.9 trend=	0.12 OH	WAVERLY
28.1	28.1	27.7	27.6	29.4 trend=	0.07 OH	NORWALK WWTP
28	28.8	28.7	28.7	30.4 trend=	0.05 OH	TIFFIN
27.5	28.2	28.1	28.6	29.8 trend=	0.01 OH	OBERLIN
28.2	28.4	29	29.8	29.5 trend=	0.01 OH	PHILO 3 SW
27.5	28.3	28.3	28.9	30.2 trend=	0.01 OH	URBANA WWTP
28	29	28.6	29	29.9 trend=	0.00 OH	KENTON
28.4	29.5	29.4	29.7	30.7 trend=	-0.01 OH	DELAWARE
28.7	29.3	29.2	29.9	30.7 trend=	-0.04 OH	CIRCLEVILLE
28.7	28.5	29.1	30.6	30.2 trend=	-0.04 OH	MC CONNELLSVILLE LK 7
27.2	28.3	28.4	27.9	29.1 trend=	-0.04 OH	WAUSEON WTP
27.8	29.5	29	28.9	30.7 trend=	-0.05 OH	BUCYRUS
27.8	28.5	28.6	28.5	29.7 trend=	-0.09 OH	FINDLAY WPCC
27.6	28.4	28.3	28.6	30 trend=	-0.15 OH	UPPER SANDUSKY
27.8	29.3	28.4	27.9	29.9 trend=	-0.23 OH	GREENVILLE WTP
33.5	39.7	37	35.2	35.7 trend=	0.39 OK	HOBART MUNI AP
33	34.2	33.7	33.4	33 trend=	0.33 OK	KENTON
33.6	37.8	33.6	34	34.4 trend=	0.24 OK	GOODWELL RSCH STN
32.5	35.6	33.2	33.1	32.7 trend=	0.21 OK	BOISE CITY 2 E
33.5	39.1	35.5	35.2	34.6 trend=	0.19 OK	CARNEGIE 5 NE
34.1	40.2	37.6	36.5	35.6 trend=	0.18 OK	ALTUS IRIG RSCH STN
33.5	38.2	33.6	33.7	33.4 trend=	0.13 OK	ANTLERS
33	39	36.1	34.9	34.3 trend=	0.11 OK	ERICK
34.2	39.2	35.5	34.7	35.5 trend=	0.11 OK	WEBBERS FALLS 5 WSW
32.8	38	34.2	34.1	33.9 trend=	0.09 OK	GEARY
33.7	39	35.2	34.7	34.6 trend=	0.09 OK	HOLDENVILLE 2SSE
33.5	39.4	37.5	35	34.8 trend=	0.09 OK	WAURIKA
33	38.5	34.8	33.9	34.1 trend=	0.08 OK	ADA
33.4	38.4	34.7	34.5	34.5 trend=	0.07 OK	OKEENE
32.9	38.2	34.3	34	33.3 trend=	0.06 OK	GUTHRIE 5S
32.8	38.9	35.1	34.1	34.3 trend=	0.06 OK	PAULS VALLEY 4 WSW
32.9	38.8	35.2	34.2	34.2 trend=	0.05 OK	ARDMORE
33	38	34	33.2	34.2 trend=	0.02 OK	POTEAU WTR WKS
32.7	37.8	33.8	33.5	33.5 trend=	0.01 OK	OKEMAH
33.6	38.4	34.7	34.8	34.6 trend=	0.00 OK	HENNESSEY 4 ESE
33.6	37.1	34.4	34.5	34.2 trend=	0.00 OK	HOOKER
33.5	38.6	34.6	34.7	34.6 trend=	0.00 OK	PERRY
32	37.2	33.1	32.9	32.4 trend=	-0.02 OK	MEEKER 5 W
33.2	38	34.1	33.8	34 trend=	-0.02 OK	OKMULGEE WTR WKS
33	38.7	35	34.9	34.3 trend=	-0.03 OK	WEATHERFORD
32.9	37.4	33.4	34	34 trend=	-0.04 OK	BEAVER
33.2	38.1	34.4	34.5	34.4 trend=	-0.04 OK	BUFFALO 2 SSW
33.2	39	36.2	34.7	34.5 trend=	-0.04 OK	LAWTON
32.8	37.8	33.9	34.1	33.9 trend=	-0.04 OK	STILLWATER 2 W
34.2	38.8	34.9	35	35 trend=	-0.05 OK	CHEROKEE
32.5	38.5	34.3	34.5	34.1 trend=	-0.05 OK	ENID
32.2	37.2	33.5	32.5	33.5 trend=	-0.05 OK	MUSKOGEE

33.4	39	35.5	34.1	34.2	trend=	-0.08	OK	DURANT
33.3	38.7	34.8	35.2	34.9	trend=	-0.08	OK	HAMMON 3 SSW
33	39.1	36.4	35.3	34.4	trend=	-0.08	OK	MANGUM
33.3	37.9	34.2	34.3	34.3	trend=	-0.10	OK	JEFFERSON
33.3	38.5	34.8	34.6	34.2	trend=	-0.11	OK	KINGFISHER
32.4	37.5	33.4	33.5	33.6	trend=	-0.15	OK	BARTLESVILLE MUNI AP
31.9	36.8	33.5	32.8	33.5	trend=	-0.15	OK	TAHLEQUAH
32	36.8	33.4	32.9	33	trend=	-0.17	OK	CLAREMORE 2 ENE
31.6	36.9	33.8	32.8	33.7	trend=	-0.18	OK	MIAMI
31.4	36.1	33.2	33.2	32.8	trend=	-0.19	OK	PAWHUSKA
31.6	37.2	33	33.8	33.7	trend=	-0.30	OK	MUTUAL
31.6	36.3	32.8	32.7	32.9	trend=	-0.35	OK	NEWKIRK 1NW
31.6	30.6	30.4	32	29.5	trend=	0.97	OR	BEND
36.1	34.9	34.8	37	35.6	trend=	0.94	OR	GRANTS PASS
32.3	31.7	31.2	31.6	31.1	trend=	0.92	OR	MCKENZIE BRG RS
34.3	32.4	32.6	32.7	31.3	trend=	0.87	OR	UNION EXP STN
30.6	30.2	29	32.2	28.2	trend=	0.82	OR	MALHEUR REFUGE HQ
32.1	31.4	30.9	32.6	29.5	trend=	0.81	OR	PAISLEY
33.3	32.4	33.6	34.8	33.7	trend=	0.80	OR	ROSEBURG KQEN
34.3	33.2	33	34	32.3	trend=	0.79	OR	DUFUR
28.6	28.9	28.9	29.9	28.6	trend=	0.79	OR	HOOD RIVER EXP STN
33.8	32.9	32.4	33.3	30.7	trend=	0.78	OR	DANNER
32.6	31.4	31.3	32.2	30.4	trend=	0.78	OR	MORO
34.2	32.9	33.3	34.9	32.5	trend=	0.77	OR	ASHLAND
37	35.7	35	35.8	33.9	trend=	0.77	OR	VALE
34.5	32.8	33.3	33.5	31.8	trend=	0.77	OR	WALLOWA
24.1	22.3	22.3	24.3	20.3	trend=	0.76	OR	CRATER LAKE NPS HQ
31.1	30.2	31.5	32.2	31.6	trend=	0.76	OR	DRAIN
33.5	31.5	32.2	32.8	30.8	trend=	0.74	OR	BAKER CITY AP
28.2	27.8	28.6	29.5	28.4	trend=	0.74	OR	CASCADIA
30.9	29.5	30.4	31.6	30.5	trend=	0.71	OR	CORVALLIS STATE UNIV
33	31.8	31.4	33	30	trend=	0.71	OR	LAKEVIEW 2 NNW
32.2	31	30.8	32.1	30.2	trend=	0.70	OR	CONDON
34.9	33.4	33.4	35.2	33	trend=	0.70	OR	KLAMATH FALLS 2 SSW
31.1	30	30.8	31.9	31	trend=	0.69	OR	MC MINNVILLE
33.7	33.1	33.6	34.9	33.5	trend=	0.69	OR	MILTON FREEWATER
33	30.7	31.2	33.5	31.9	trend=	0.67	OR	PROSPECT 2 SW
35.9	34.5	34.6	35.4	33.8	trend=	0.66	OR	HERMISTON 1 SE
27.6	27	27.8	28.4	27.7	trend=	0.66	OR	THREE LYNX
29.4	28	29.5	30.9	30.1	trend=	0.65	OR	COTTAGE GROVE 1 NNE
33	32.5	31.5	33.3	30.3	trend=	0.65	OR	FREMONT 5 NW
33.1	31.8	33	34.6	33.9	trend=	0.65	OR	RIDDLE
35.5	34.1	34.1	35.1	33.3	trend=	0.57	OR	PILOT ROCK 1 SE
25.4	25.8	25.9	26.5	25.4	trend=	0.47	OR	HEADWORKS PORTLAND WTR
33.6	32.5	32.2	33.5	31.4	trend=	0.46	OR	PRINEVILLE
28.9	28.8	29.4	29.9	28.5	trend=	0.42	OR	FOREST GROVE
33	31.8	31.1	32.3	30.8	trend=	0.40	OR	HEPPNER

17.9	17.6	18.3	19.2	17.9 trend=	0.23 OR	BROOKINGS 2 SE
19.6	21.2	21.5	22.3	20.4 trend=	0.21 OR	ASTORIA AP PORT OF
19.3	20.4	21	20.6	19.3 trend=	0.18 OR	TILLAMOOK 1 W
18.6	19	19.5	19.4	17.9 trend=	0.14 OR	NORTH BEND FCWOS
17.8	18.1	18.6	18.8	18.1 trend=	0.09 OR	NEWPORT
25.5	27.8	26.7	27.4	28.1 trend=	0.63 PA	WELLSBORO 4 SW
25.6	26.8	25.9	26.5	27.3 trend=	0.58 PA	PLEASANT MT 1 W
29.1	31.6	29.8	30.9	30.5 trend=	0.56 PA	WILLIAMSPORT RGNL AP
29.4	30.9	30.1	31.2	31.8 trend=	0.51 PA	NEW CASTLE 1 N
29.1	31	29.6	30.7	30.6 trend=	0.48 PA	ALLENTOWN AP
30.4	31.3	31	31.8	31.4 trend=	0.46 PA	READING 4 NNW
29.7	30.8	29.8	30.8	29.6 trend=	0.46 PA	SELINGSGROVE 2 S
30.4	31.3	31	31.5	31.5 trend=	0.43 PA	WEST CHESTER 2 NW
25.9	28.3	27.1	27.9	28.2 trend=	0.42 PA	MONTROSE
30.6	30.3	31.4	32.2	31.5 trend=	0.41 PA	EISENHOWER NHS
28.7	29.8	29.6	30.5	28 trend=	0.38 PA	LEBANON 2 W
30.3	30.4	31.3	31.9	31.9 trend=	0.37 PA	CHAMBERSBURG 1 ESE
27.6	29	29.2	29.9	30.1 trend=	0.34 PA	TOWANDA 1 S
30.8	31.2	31.8	32.7	31.6 trend=	0.34 PA	YORK 3 SSW PUMP
26.8	28.3	27.5	28.5	29.7 trend=	0.32 PA	RIDGWAY
27.2	28.7	28.5	28.9	30.4 trend=	0.31 PA	FRANKLIN
30.9	32.9	31.9	32.9	32.6 trend=	0.31 PA	PALMERTON
27.8	29	28.5	29.6	30.3 trend=	0.30 PA	GREENVILLE 2 NE
27.7	28.6	28	28.8	29.2 trend=	0.28 PA	STATE COLLEGE
25.5	27.1	26.6	27.6	28 trend=	0.26 PA	ERIE WSO AP
30.7	31.3	30.8	32.7	32.3 trend=	0.26 PA	JOHNSTOWN
29.3	28.9	29.5	30.9	30.9 trend=	0.25 PA	UNIONTOWN 1 NE
26.5	28.6	27.5	28.4	29 trend=	0.14 PA	WARREN
28	30	29.1	29.8	30.4 trend=	0.11 PA	STROUDSBURG
26.8	28.8	28.6	28	28.6 trend=	0.35 RI	KINGSTON
25	27.4	27	26.8	27.3 trend=	0.30 RI	BLOCK ISLAND STATE AP
26.5	30.2	28.5	28.3	29.3 trend=	0.30 RI	PROVIDENCE WSO AP
33.2	33.5	34	34.7	35.4 trend=	0.39 SC	COLUMBIA UNIV OF SC
31.5	33.3	33.7	34.3	34.2 trend=	0.33 SC	YEMASSEE
31.3	32.2	32.3	32.5	34.5 trend=	0.28 SC	SANTUCK
30.5	31.3	31.5	31.2	32.8 trend=	0.26 SC	WALHALLA
30.5	31.7	31.8	31.9	33.4 trend=	0.25 SC	WINNSBORO
30.7	32.9	32.4	32.7	34.2 trend=	0.21 SC	CLEMSON UNIV
32.2	32.4	33.5	33.7	34.1 trend=	0.21 SC	DARLINGTON
32.2	33.1	33.7	33.8	34.8 trend=	0.20 SC	BLACKVILLE 3 W
31.1	33.1	32.9	33	34.4 trend=	0.19 SC	ANDERSON
31.1	32	32.3	33.2	34.3 trend=	0.18 SC	CALHOUN FALLS
31.3	30.7	31.9	30.9	31.3 trend=	0.17 SC	CHARLESTON CITY
31.9	32.9	33.9	33.5	33.5 trend=	0.17 SC	SUMMERVILLE 4W
31.5	32.7	32.6	33.5	34.5 trend=	0.15 SC	LAURENS
31.3	31.8	32.5	33.6	33.8 trend=	0.14 SC	CHERAW
31.1	31.5	32.4	32.8	33.5 trend=	0.13 SC	CAMDEN 3 W

31	31.6	32.3	32.6	33.6	trend=	0.10	SC	LITTLE MTN
31.7	32.6	33.3	34.4	34.6	trend=	0.10	SC	NEWBERRY
31.7	32.7	33.3	34.1	34.3	trend=	0.09	SC	BEAUFORT WWTP
31.8	32.7	33.4	32.8	33.5	trend=	0.09	SC	CONWAY
31.8	32.7	33.2	33.5	34.4	trend=	0.09	SC	KERSHAW 1SW
32.4	32.9	33.7	34	34.4	trend=	0.08	SC	AIKEN 5SE
30.6	31.6	32.3	31.5	32.7	trend=	0.04	SC	GEORGETOWN 2 E
32.5	33.4	33.7	34.1	35.8	trend=	0.04	SC	GREENWOOD
31.9	32.8	32.8	33.4	34.5	trend=	0.02	SC	GRNVL SPART INTL AP
32.4	32.9	33.9	34.6	35.3	trend=	0.01	SC	SALUDA
32.6	33.2	33.8	34.3	34.5	trend=	0.00	SC	KINGSTREE
31.6	31.5	32.5	32.4	31.5	trend=	0.00	SC	ORANGEBURG 2
32.2	32.5	33.4	34.3	34.3	trend=	-0.03	SC	SUMTER
31.4	32.5	32.4	32.4	34.2	trend=	-0.08	SC	WINTHROP UNIV
31.5	31.9	28.4	31.4	29.5	trend=	0.16	SD	RAPID CITY 4NW
32.4	31.4	29.3	30	31.2	trend=	0.07	SD	MURDO
33.6	32.5	30.3	32.8	31.3	trend=	0.04	SD	COTTONWOOD 2 E
29.9	30.8	28.8	28.5	29	trend=	0.01	SD	ALEXANDRIA
31.7	32.7	29.5	33.4	33.1	trend=	-0.07	SD	HOT SPRINGS
31.6	31.8	29.8	29.4	31.8	trend=	-0.09	SD	ACADEMY 2NE
29.2	30	27.6	28	29.3	trend=	-0.09	SD	CANTON
33	32.2	30.8	30.4	32.5	trend=	-0.12	SD	KENNEBEC
34.6	34.2	31.2	31.7	31.1	trend=	-0.16	SD	DUPREE
33.4	31.8	31	30.8	31.3	trend=	-0.18	SD	PIERRE RGNL AP
30.1	30.3	28.8	28.6	29.5	trend=	-0.22	SD	FORESTBURG 3 NE
30	27.5	25.7	27	27.4	trend=	-0.23	SD	EUREKA
31.3	30.1	29.3	29.8	28.1	trend=	-0.25	SD	ABERDEEN RGNL AP
31.4	30.8	29.3	29.1	30.7	trend=	-0.28	SD	GANN VALLEY 4NW
30	31.9	28.7	28.6	29.2	trend=	-0.28	SD	MENNO
33.7	32	30.6	30.5	31.2	trend=	-0.28	SD	OAHE DAM
30.4	29.8	28.8	28.7	29	trend=	-0.30	SD	MILBANK 4 NW
29	28.2	27.7	27.6	27.1	trend=	-0.30	SD	WATERTOWN RGNL AP
29.1	28.4	27.4	27.3	27.7	trend=	-0.31	SD	CLARK
32.7	32.2	30.4	30.7	30.7	trend=	-0.39	SD	WOOD
30.5	30	28.6	29.7	28.7	trend=	-0.40	SD	FAULKTON 1 NW
28.7	30.9	28.1	28.8	29.2	trend=	-0.42	SD	VERMILLION 2 SE
30.5	29.6	28	28.5	28.6	trend=	-0.44	SD	MELLETTE 4 W
31	30.2	28.4	28.3	30	trend=	-0.49	SD	HIGHMORE 1 W
30.1	29.6	28.1	27.8	28.7	trend=	-0.53	SD	HOWARD
27	28.7	27.3	28.2	29.3	trend=	0.22	TN	CROSSVILLE ED & RESEARCH
31.4	34.5	32.6	33.1	33.9	trend=	0.17	TN	LEWISBURG EXP STN
32	35.3	32.6	32.2	33.8	trend=	0.07	TN	JACKSON EXP STN
30.1	32.7	30.8	31.5	32.2	trend=	0.04	TN	TULLAHOMA
32.5	35.2	33	33	35	trend=	0.03	TN	CLARKSVILLE WWTP
30.3	32.1	30.4	31.3	32	trend=	0.01	TN	MC MINNVILLE
31	32.2	30.7	31.9	32.6	trend=	0.01	TN	NEWPORT 1 NW
30.5	30.6	29.6	30.7	31.8	trend=	0.01	TN	ROGERSVILLE 1 NE

31.2	33.2	31.7	32.4	32.7	trend=	-0.02	TN	MURFREESBORO 5 N
29.8	31.7	30.5	31.4	32.3	trend=	-0.04	TN	COPPERHILL
31	33.3	31.2	29.8	31.8	trend=	-0.05	TN	DOVER 1 W
31.2	33.4	31.6	31.2	33.1	trend=	-0.11	TN	COVINGTON 3 SW
30.7	33.2	32	32	32.5	trend=	-0.11	TN	WAYNESBORO
29.9	31.9	30.1	30.8	32	trend=	-0.18	TN	DICKSON
31.2	33	31.1	31.2	33.1	trend=	-0.30	TN	UNION CITY
32.5	38.7	38.2	34.6	34.3	trend=	0.49	TX	SAN ANTONIO INTL AP
31.4	34.7	36.8	33.5	32	trend=	0.45	TX	ALPINE
33.3	39	38.6	36	33.1	trend=	0.43	TX	BALLINGER 2 NW
34.7	38.1	40.6	37.1	36	trend=	0.42	TX	EL PASO AP
33.8	40	39.6	35.2	34.5	trend=	0.41	TX	LULING
35.9	39.4	40.6	37.7	36.1	trend=	0.37	TX	FT STOCKTON
31.1	37.2	34.2	34	32.3	trend=	0.35	TX	PLAINVIEW
32.8	39.1	37.4	36.2	34	trend=	0.34	TX	SNYDER
32.9	35.3	36	33	34.1	trend=	0.33	TX	LIBERTY
34.7	39.1	40.2	37.2	35.8	trend=	0.33	TX	PECOS
32	36	38.2	34	34.6	trend=	0.31	TX	BEEVILLE 5 NE
33.5	39.5	38.3	35.1	31.8	trend=	0.31	TX	BROWNWOOD 2ENE
34	39.3	38.7	35.9	34.4	trend=	0.31	TX	EAGLE PASS 3N
32.5	38.5	37.4	34.8	32.8	trend=	0.30	TX	DUBLIN 2SE
34.4	39.8	39.2	36.2	35.7	trend=	0.30	TX	ENCINAL
36.2	42.1	41.8	38.1	37.6	trend=	0.29	TX	CATARINA
34.2	38.7	38.9	35.5	36.5	trend=	0.28	TX	FALFURRIAS
33.2	38.5	38.3	33.9	34.3	trend=	0.27	TX	HALLETTSVILLE 2 N
30.3	36.4	35.4	33.2	32.1	trend=	0.26	TX	MULESHOE #1
32	37	37.4	34.4	33.5	trend=	0.26	TX	SEMINOLE
31.8	37.4	34.9	34.2	33	trend=	0.24	TX	MIAMI
32.9	36.7	38.2	35.1	33.5	trend=	0.23	TX	BALMORHEA
32.6	38.3	38.6	34.3	33.6	trend=	0.23	TX	LAMPASAS
32.2	35.7	36.3	33	33.5	trend=	0.22	TX	DANEVANG 1 W
33.3	35.2	36.9	34.4	34.7	trend=	0.21	TX	CORPUS CHRISTI AP
32.6	36.5	34.5	34.4	32.9	trend=	0.21	TX	STRATFORD
32	39	35.5	35.2	33.6	trend=	0.17	TX	CROSBYTON
32.5	36.6	35.1	31.9	34.1	trend=	0.17	TX	MARSHALL
31.3	37	36.7	32.7	32.6	trend=	0.15	TX	BLANCO
35	38.7	38.4	35.8	34.8	trend=	0.15	TX	MCCAMEY
33.2	39.4	37.9	36.1	33.8	trend=	0.14	TX	ALBANY
34.2	38.2	38.7	35.2	36.2	trend=	0.14	TX	ALICE
34.4	39.4	37.6	35.8	34.8	trend=	0.13	TX	HASKELL
31.8	36.2	34.6	34.6	32.1	trend=	0.12	TX	BOYS RANCH
32.6	38.4	38.3	34	33.7	trend=	0.12	TX	NEW BRAUNFELS
31.7	37.4	36.9	32.7	33.6	trend=	0.11	TX	FLATONIA
33.3	38.9	39.3	35	33.3	trend=	0.11	TX	LLANO
31.8	36.7	37.5	33.8	32.6	trend=	0.10	TX	BOERNE
32	39.1	36.9	33.8	33.6	trend=	0.10	TX	CORSICANA
31.9	37.6	36.6	33	32.9	trend=	0.09	TX	TEMPLE

32.6	38.3	37	33.4	33.8 trend=	0.07 TX	MEXIA
31.8	37.3	36.4	33.6	34.6 trend=	0.06 TX	BRENNHAM
35.2	40.7	36.6	35.5	36.5 trend=	0.04 TX	GREENVILLE KGV L RADIO
34.3	39	34.2	34.3	35.2 trend=	0.01 TX	PARIS
31.9	38.9	35.2	33.4	34 trend=	0.00 TX	GAINESVILLE 5 ENE
33.6	39.9	37	36.5	35.1 trend=	-0.01 TX	QUANAH 2 SW
34.6	39.2	39	35.8	36.7 trend=	-0.01 TX	RIO GRANDE CITY
32.6	38.7	37.8	34.2	33.5 trend=	-0.01 TX	WEATHERFORD
32.5	37	33	31.8	33.6 trend=	-0.02 TX	CLARKSVILLE 2NE
29.9	27.2	29.5	28.3	27.5 trend=	0.94 UT	SALINA 24 E
33.1	33	32.5	32.2	31.9 trend=	0.87 UT	SNAKE CREEK POWERHOUSE
35.4	35.5	34.7	34.3	34.3 trend=	0.85 UT	TOOELE
34.3	33.2	35.6	35	34.5 trend=	0.83 UT	ESCALANTE
36.1	36.1	35.5	35.4	34.6 trend=	0.81 UT	WENDOVER AP AWOS
35	34.9	36.5	36.9	35.1 trend=	0.74 UT	KANAB
35.8	35.2	35.1	35.1	34.7 trend=	0.73 UT	FT DUCHESNE
34	33.8	32.8	31.6	32.2 trend=	0.73 UT	LOGAN UTAH ST UNIV
35.7	34.9	34.4	34.4	33.8 trend=	0.72 UT	MORGAN POWER & LIGHT
39.9	40	42	42.3	39.3 trend=	0.71 UT	ST GEORGE
30	30.1	31.5	32	30.7 trend=	0.64 UT	ALTON
35.4	35.3	34.6	34.1	34.3 trend=	0.64 UT	FARMINGTON 3 NW
36.7	35.9	38.2	37	36.3 trend=	0.63 UT	BLUFF
35.9	35.8	35.2	35.3	35 trend=	0.63 UT	OGDEN PIONEER P H
31	30.7	30.4	31.1	30.4 trend=	0.63 UT	WOODRUFF
38.1	37.4	37.5	37.6	36 trend=	0.60 UT	DESERET
32.6	32.5	32.5	32	30.8 trend=	0.58 UT	HEBER
34.8	34.9	34.8	34	33.6 trend=	0.57 UT	SCIPIO
38.9	37.9	39.2	40.2	37.7 trend=	0.56 UT	ZION NP
33.9	32.9	33.2	32.4	31.9 trend=	0.54 UT	VERNAL 2SW
32.5	31.5	33.7	33.2	32.2 trend=	0.53 UT	BLANDING
35.4	35.1	34.3	34.3	34 trend=	0.53 UT	CORINNE
37.2	36.3	38.3	38.2	36.7 trend=	0.53 UT	MOAB
34.6	34.5	34.8	34.7	33.9 trend=	0.52 UT	SPANISH FORK PWR HOUSE
32.7	32.6	34.5	34.4	32.5 trend=	0.51 UT	PAROWAN PWR
32.4	32.9	32.6	32.1	31.4 trend=	0.49 UT	MANTI
38.7	37.6	39.4	38.3	38.2 trend=	0.47 UT	GREEN RIVER AVIATION
38.4	37.9	39.2	38.5	37.6 trend=	0.46 UT	HANKSVILLE
24.9	25	24.8	24.6	24.2 trend=	0.46 UT	SCOFIELD-SKYLINE MINE
33.7	33.4	33.3	33.2	30.4 trend=	0.46 UT	UTAH LAKE LEHI
34.1	34.5	34.1	33.9	32.9 trend=	0.45 UT	LEVAN
35.5	34.8	35.9	35.2	35.1 trend=	0.45 UT	THOMPSON
29.1	29	31.2	30.9	29.6 trend=	0.44 UT	PANGUITCH
34.2	34.3	35.7	35.8	33.6 trend=	0.43 UT	MODENA
28.4	28.6	27.7	27.5	27.6 trend=	0.38 UT	LAKETOWN
34.9	35.2	34.2	34.1	34 trend=	0.35 UT	NEPHI
32.4	33.2	34.5	33.8	32.6 trend=	0.33 UT	RICHFIELD RADIO KSVC
31.8	31.7	32.3	31.8	31.6 trend=	0.29 UT	DUCHESNE

35.2	34.5	35.2	34	32.9 trend=	0.29 UT	FILLMORE
31.9	31.7	33.2	32.9	31.7 trend=	0.06 UT	MARYSVALE
30.4	29.7	31.4	31.6	31.3 trend=	0.55 VA	LINCOLN
34	33.8	34	34.8	34.6 trend=	0.41 VA	BREMO BLUFF
32.6	30.7	31.4	32.5	32.3 trend=	0.41 VA	CHARLOTTESVILLE 2W
28.6	29.6	29.2	29.6	29.8 trend=	0.40 VA	BLACKSBURG NWSO
31.4	32	32	31.4	31.9 trend=	0.38 VA	NORFOLK INTL AP
30.7	29.2	30.1	31.5	31.3 trend=	0.31 VA	DALE ENTERPRISE
26.2	26.3	26.3	27.3	27.4 trend=	0.29 VA	BURKES GARDEN
33.1	32.8	33.3	33.7	33.8 trend=	0.28 VA	FREDERICKSBURG NP
30.8	32	31.7	30.9	32.2 trend=	0.27 VA	WILLIAMSBURG 2 N
32.4	29.9	30.9	32	31.6 trend=	0.27 VA	WOODSTOCK 2 NE
31.3	31.1	31	31.8	32.1 trend=	0.25 VA	LEXINGTON
30.6	31.2	31.8	30.9	31.3 trend=	0.24 VA	HOPEWELL
31.1	31.4	31.5	31.6	32.3 trend=	0.18 VA	FARMVILLE 2 N
31.8	31.1	32	32.4	32.5 trend=	0.17 VA	PIEDMONT RSCH STN
30.5	30.6	30.5	30.9	31.1 trend=	0.17 VA	ROCKY MT
32	32.6	32.3	32.6	33.4 trend=	0.14 VA	DANVILLE
30.9	29.7	30.3	31.6	31.2 trend=	0.10 VA	STAUNTON WATER TRMTMT PLT
28.5	28.3	28.4	29.1	30.2 trend=	0.02 VA	HOT SPRINGS
30.7	30.9	30.7	31.8	32.1 trend=	0.01 VA	PENNINGTON GAP
25.2	28.6	28.3	29.7	29.2 trend=	0.30 VT	SOUTH HERO
22.4	25.2	26.1	27.1	26.1 trend=	0.30 VT	SOUTH LINCOLN
25.4	29.2	28.7	29.9	29.7 trend=	0.27 VT	CAVENDISH
24.8	28.3	28.4	29	28.8 trend=	0.26 VT	CORNWALL
24.2	27.9	27.7	28.7	28.3 trend=	0.24 VT	CHELSEA
24.2	26.7	26.9	28.7	28.2 trend=	0.23 VT	ENOSBURG FALLS
25.4	28.9	28.8	29.6	29.1 trend=	0.20 VT	BURLINGTON WSO AP
24.6	28	28.4	28.6	28.9 trend=	0.10 VT	SAINT JOHNSBURY
34.9	31.8	32.2	33.9	31.2 trend=	0.94 WA	COLVILLE
32.3	30.1	30.4	31.7	29.5 trend=	0.79 WA	SAINT JOHN
34.3	33.2	32.7	33.8	32.3 trend=	0.76 WA	SUNNYSIDE
27	25.6	26.2	25.4	24.8 trend=	0.74 WA	LONGMIRE RAINIER NPS
27.3	27.3	27.5	28.7	26.6 trend=	0.72 WA	VANCOUVER 4 NNE
35.4	31.5	32.6	33.7	31.4 trend=	0.69 WA	NORTHPORT
34.4	33.1	33.1	34.1	32.7 trend=	0.69 WA	WALLA WALLA FAA AP
35.8	34.9	34.8	35.3	34.4 trend=	0.67 WA	KENNEWICK
34.4	32.5	31.8	33.3	31.4 trend=	0.67 WA	ODESSA
33	30.9	31.3	33.7	30 trend=	0.65 WA	SPOKANE INTL AP
33.2	28.8	32	31.9	30.7 trend=	0.64 WA	PULLMAN 2 NW
33	30.7	30.1	32	31.3 trend=	0.64 WA	STEHEKIN 4 NW
27.5	28.3	28.8	29.6	28.1 trend=	0.62 WA	LONGVIEW
33.7	31.9	31.5	32.6	31.2 trend=	0.61 WA	GOLDENDALE
23.9	23.5	24.2	24.2	23.5 trend=	0.61 WA	PORT TOWNSEND
32.4	30	29.7	31.6	29.7 trend=	0.60 WA	CONCONULLY
26.1	25.9	26.3	26.8	26.2 trend=	0.60 WA	SNOQUALMIE FALLS
23.3	23	23.8	23.7	22.3 trend=	0.59 WA	BELLINGHAM 3 SSW

32.2	29.5	30.1	31.7	29.6	trend=	0.58	WA	DAVENPORT
33.2	31.5	30.6	32.8	30.6	trend=	0.58	WA	WINTHROP 1 WSW
33.1	30.8	30.7	32.3	30.4	trend=	0.57	WA	WILBUR
33.3	32	32.1	32.8	31.8	trend=	0.56	WA	DAYTON 1 WSW
30	29	29	30.5	28.4	trend=	0.55	WA	CLE ELUM
32.5	31.3	30.8	32.1	30.7	trend=	0.55	WA	ELLENSBURG
26.2	26.6	26.6	27.5	26.3	trend=	0.55	WA	SEATTLE URBAN SITE
25.2	24.8	25.8	25.7	25.4	trend=	0.54	WA	SEDRO WOOLLEY
25.3	25.1	25.8	26.2	25.5	trend=	0.52	WA	CUSHMAN POWERHOUSE 2
22.1	22.2	22.6	23.4	22.4	trend=	0.50	WA	CEDAR LAKE
25.9	26.2	26.2	27	26.2	trend=	0.49	WA	BUCKLEY 1 NE
26.5	26.8	26.8	27.8	26.5	trend=	0.49	WA	CENTRALIA
26	25.6	25.1	25.5	25.7	trend=	0.49	WA	CLEARBROOK
22.5	21.7	22.1	22.4	20.9	trend=	0.48	WA	OLGA 2 SE
31.8	29.8	29.3	31.1	29	trend=	0.48	WA	WATERVILLE
33.5	31.9	32.1	33.4	31.2	trend=	0.45	WA	RITZVILLE 1 SSE
35.2	32.7	33.3	35.2	33.1	trend=	0.43	WA	WENATCHEE
33	31.1	31.7	32.7	31	trend=	0.41	WA	POMEROY
22.8	23.4	24	24.2	22.7	trend=	0.40	WA	FORKS 1 E
20.1	20	20.8	20.7	19.3	trend=	0.36	WA	PORT ANGELES
23.7	23.2	23.9	23.7	-999	trend=	0.35	WA	BLAINE
25.5	25.7	26	26.7	25.6	trend=	0.31	WA	MC MILLIN RSVR
25	25	25.6	25.7	25.3	trend=	0.26	WA	EVERETT
18.6	19.6	20	19.8	18.7	trend=	0.20	WA	LONG BEACH EXP STN
20.9	23.1	23.5	23.8	22	trend=	0.15	WA	RAYMOND 2 S
19.6	21.3	21.8	21.2	20.2	trend=	0.06	WA	ABERDEEN
27.2	27	27.1	26.3	27.8	trend=	0.12	WI	BOWLER
27.9	26.4	26.1	26.9	26.7	trend=	0.09	WI	ASHLAND EXP FARM
27.3	27.1	27.8	26.7	28	trend=	0.06	WI	HANCOCK EXP FARM
25.3	24.3	23.8	23.8	25.2	trend=	0.05	WI	MINOCQUA
27	27	26.7	26	27.4	trend=	0.03	WI	OCONTO 4 W
28.3	28.3	28.3	27.9	30.1	trend=	0.02	WI	BRODHEAD
27.9	28.1	28.9	27.6	29.2	trend=	0.00	WI	MILWAUKEE MT MARY COL
25.3	26.4	25.9	25.4	27.1	trend=	-0.01	WI	RACINE
27.1	25.9	25.9	25.4	26.7	trend=	-0.02	WI	MEDFORD
28	28.5	28.7	27	28.7	trend=	-0.03	WI	DARLINGTON
27.8	27.8	28.6	27.3	28.7	trend=	-0.04	WI	OSHKOSH
27.4	27.9	28.3	27.1	28.8	trend=	-0.04	WI	PORTAGE
27.9	27.3	27.2	26.5	27.1	trend=	-0.04	WI	SPOONER AG RES STN
26.6	25.9	26.8	25.7	27.3	trend=	-0.05	WI	MARSHFIELD EXP FARM
26.7	26.6	27.1	25.9	27.9	trend=	-0.08	WI	FOND DU LAC
27.7	27.2	27.3	26.6	27.7	trend=	-0.12	WI	STANLEY
27.1	27.1	28.3	27.1	28.6	trend=	-0.13	WI	NEW LONDON
27.3	26.6	27.4	26.7	27.9	trend=	-0.14	WI	NEILLSVILLE 3 SW
27.3	27.4	27.4	26.2	27.9	trend=	-0.21	WI	LANCASTER 4 WSW
28	27.4	28.3	26.4	28.6	trend=	-0.22	WI	WATERTOWN
24.4	24.5	25.2	23.8	24.2	trend=	-0.25	WI	MANITOWOC

26.2	26.5	27	25.9	27.5 trend=	-0.25 WI VIROQUA
28.9	28.6	28.6	27.6	29.3 trend=	-0.32 WI PRAIRIE DU CHIEN
30	29.9	30.4	32	32 trend=	0.42 WV MARTINSBURG E WV RGNL
28.6	28.8	29.1	29.9	30.9 trend=	0.39 WV LEWISBURG 3 N
31.6	32	32.5	32.6	34.5 trend=	0.33 WV WILLIAMSON
28.9	28.5	29.1	30.6	30.5 trend=	0.29 WV MANNINGTON 8 WNW
28.5	28.3	29.1	30.5	30 trend=	0.23 WV PARSONS 1 NE
25.3	25.2	25.5	26.8	27 trend=	0.21 WV PICKENS 2 N
29.9	29.7	29.8	31.7	31.5 trend=	0.20 WV GLENVILLE
29.3	29	29.5	31.4	31.2 trend=	0.19 WV BUCKHANNON
29.6	29.1	29.9	31.7	30.9 trend=	0.18 WV CAIRO
28.7	29.1	29.2	30.5	30.6 trend=	0.17 WV WELLSBURG WTR TRMT PL
28.5	28.5	28.8	29.8	30.3 trend=	0.16 WV PINEVILLE
28.5	28	28.6	31.1	30 trend=	0.02 WV SPENCER
29.4	29.5	30.4	31.9	32 trend=	-0.04 WV WINFIELD LOCKS
31.9	30.9	28.2	30	29.1 trend=	1.31 WY CODY
32	31	30.3	30	30.3 trend=	0.83 WY ROCK SPRINGS AP
25.6	24.5	23.3	22.9	23.5 trend=	0.71 WY LAKE YELLOWSTONE
35.9	35.3	33	34.2	34.4 trend=	0.66 WY WORLAND
30.9	29.8	29	28.5	28.5 trend=	0.66 WY YELLOWSTONE PK MAMMOTH
32.8	32	31.7	31.6	31.9 trend=	0.63 WY GREEN RIVER
27	26.7	24.7	24.6	25.5 trend=	0.62 WY MORAN 5 WNW
28.9	28	26.3	26.2	27.2 trend=	0.61 WY PINEDALE
30.2	29.9	28.5	28.3	27.9 trend=	0.59 WY EVANSTON 1 E
28.8	28.9	27.1	26.7	27.3 trend=	0.58 WY ALTA 1 NNW
30.6	31	28.1	30	28.8 trend=	0.47 WY CHEYENNE WSFO AP
34.5	33.4	29.4	31.8	30.6 trend=	0.46 WY SHERIDAN FLD STN
32.3	31.7	28.9	30.8	29.4 trend=	0.43 WY POWELL FLD STN
27.6	28.5	26.5	26.7	26.2 trend=	0.37 WY DUBOIS
29.4	29.6	27.7	28.9	28.5 trend=	0.34 WY LARAMIE RGNL AP
35.4	35.1	32.7	33.5	33.7 trend=	0.30 WY BASIN
29.8	29.2	28.1	28	29.4 trend=	0.30 WY SARATOGA
32.7	32.3	29.9	30.8	31.5 trend=	0.29 WY DIVERSION DAM
33.8	33.7	30.5	33.3	32.3 trend=	0.29 WY TORRINGTON EXP FARM
31.9	32.1	29.4	31.6	31.3 trend=	0.28 WY CHUGWATER
33.3	32.4	29.5	31.7	29.9 trend=	0.23 WY COLONY
31.3	30.8	27.8	30.7	30.2 trend=	0.23 WY LUSK 2 SW
32.9	32.6	30.5	31.6	32 trend=	0.21 WY RIVERTON
33.1	32	30.1	31.4	31.7 trend=	0.19 WY MIDWEST
32	31.9	28.7	31.8	31.4 trend=	0.19 WY PINE BLUFFS 5W
31.7	31	29.3	29.8	30.7 trend=	0.17 WY PAVILLION
32.2	32.7	29.2	32	31.2 trend=	0.16 WY WHEATLAND 4 N
29	29.2	27.1	28.4	28.6 trend=	0.15 WY BATES CREEK #2
32.5	32.4	29.5	32.8	31.2 trend=	-0.10 WY NEWCASTLE

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Impacts of Carbon Dioxide Emissions on the U.S. Climate

Climate Working Group
United States Department of Energy
May 27, 2025

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Impacts of Carbon Dioxide Emissions on the U.S. Climate

Report to U.S. Energy Secretary Christopher Wright

May 27, 2025

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Suggested citation:

Climate Working Group (2025) Impacts of Carbon Dioxide Emissions on the U.S. Climate. Washington DC: Department of Energy, May 27, 2025.

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SECRETARY'S FOREWORD

Energy, Integrity, and the Power of Human Potential

Over my lifetime, I've had the privilege of working as an energy entrepreneur across a range of fields—nuclear, geothermal, natural gas, and more—and I now serve as Secretary of Energy under President Donald Trump. But above all, I'm a physical scientist who sees modern energy as nothing short of miraculous. It powers every aspect of modern life, drives every industry, and has made America an energy powerhouse with the ability to fuel global progress.

The rise of human flourishing over the past two centuries is a story worth celebrating. Yet we are told—relentlessly—that the very energy systems that enabled this progress now pose an existential threat. Hydrocarbon-based fuels, the argument goes, must be rapidly abandoned or else we risk planetary ruin.

That view demands scrutiny. That's why I commissioned this report: to encourage a more thoughtful and science-based conversation about climate change and energy. With my technical background, I've reviewed reports from the Intergovernmental Panel on Climate Change, the U.S. government's assessments, and the academic literature. I've also engaged with many climate scientists, including the authors of this report.

What I've found is that media coverage often distorts the science. Many people—even well-meaning ones—walk away with a view of climate change that is exaggerated or incomplete. To provide clarity, I asked a diverse team of independent experts to summarize the current state of climate science, with a focus on how it relates to the United States.

I didn't select these authors because we always agree—far from it. In fact, they may not always agree with each other. But I chose them for their rigor, honesty, and willingness to elevate the debate. I exerted no control over their conclusions. What you'll read are their words, drawn from the best available data and scientific assessments.

I've reviewed the report carefully and challenged its findings. I believe it faithfully represents the state of climate science today. Still, many readers may be surprised by its conclusions—which differ in important ways from the mainstream narrative. That's a sign of how far the public conversation has drifted from the science itself.

To correct course, we need open, respectful, and informed debate. That's why I'm inviting public comment on this report. Honest scrutiny and scientific transparency should be at the heart of our policymaking.

Climate change is real, and it deserves attention. But it is not the greatest threat facing humanity. That distinction belongs to global energy poverty. As someone who values data, I know that improving the human condition depends on expanding access to reliable, affordable energy. Climate change is a challenge—not a catastrophe. But misguided policies based on fear rather than facts could truly endanger human well-being.

We stand at the threshold of a new era of energy leadership. If we empower innovation rather than restrain it, America can lead the world in providing cleaner, more abundant energy—lifting billions out of poverty, strengthening our economy, and improving our environment along the way.

EXECUTIVE SUMMARY

This report reviews scientific certainties and uncertainties in how anthropogenic carbon dioxide (CO₂) emissions have affected, or will affect, the Nation's climate, extreme weather events, and selected metrics of societal well-being. Those emissions are increasing the concentration of CO₂ in the atmosphere through a complex and variable carbon cycle, where some portion of the additional CO₂ persists in the atmosphere for centuries.

Elevated concentrations of CO₂ directly enhance plant growth, globally “greening” the planet and increasing agricultural productivity. They also make the oceans less alkaline (lower the pH) by an amount within the range of natural variability. That is possibly detrimental to coral reefs, although the recent rebound of the Great Barrier Reef suggests otherwise.

Carbon dioxide also acts as a greenhouse gas, exerting a warming influence on climate and weather. Scenarios of future CO₂ emissions have tended to overestimate observed levels.

The world's several dozen detailed climate models offer little guidance on how the climate responds to elevated CO₂, with the average surface warming under a doubling of the CO₂ concentration ranging from 1.8°C to 5.7°C, a larger range than data driven estimates that suggest a response in the range 1.8°C to 2.7°C.

Detailed climate models generally run “hot” in their description of the climate of the past few decades - too much warming at the surface and too much amplification of warming in the lower- and mid-troposphere. They also produce too much stratospheric cooling, too much snow loss, and too much warming in the U.S. Corn Belt. The combination of overly sensitive models and extreme scenarios for future emissions has exaggerated projections of future warming.

Most extreme weather events do not show long-term trends. Claims of increased frequency or intensity of hurricanes, tornadoes, floods, and droughts are not supported by historical data. Additionally, forest management practices are often overlooked in assessing wildfire activity. Global sea level has risen approximately 8 inches since 1900, but there are significant regional variations in tide gauge data driven primarily by local land subsidence; U.S. tide gauge measurements in aggregate show no obvious acceleration in sea level rise beyond the historical average rate.

The attribution of particular climate changes or extreme weather events to human CO₂ emissions is challenged by natural climate variability, data limitations, and inherent model deficiencies. Moreover, solar activity's contribution to the late 20th century warming might be underestimated.

Both models and experience suggest that CO₂-induced warming might be less damaging economically than commonly believed, and excessively aggressive mitigation policies could prove more detrimental than beneficial. Social Cost of Carbon estimates, which attempt to quantify the economic damage of CO₂ emissions, are highly sensitive to their underlying assumptions and so provide limited independent information.

The report supports a more nuanced and evidence-based approach for informing climate policy. The risks and benefits of a climate changing under both natural and human influences must be weighed against the costs, efficacy, and collateral impacts of any “climate action”, considering the nation's need for reliable and affordable energy with minimal local pollution. Beyond continuing precise, un-interrupted observations of the global climate system, it will be important to make realistic assumptions about future emissions, re-evaluate climate models to address biases and uncertainties, and clearly acknowledge the limitations of extreme event attribution studies. An approach that acknowledges both the potential risks and benefits of CO₂, rather than relying on flawed models and extreme scenarios, is essential for informed and effective decision-making.

PREFACE

This document originated in late March 2025 when Secretary Wright assembled an independent group to write a report on issues in climate science relevant for energy policymaking, with particular focus on the question of whether carbon dioxide emissions endanger the U.S. public. We agreed to undertake the work on the condition that there would be no editorial oversight by the Secretary, the Department of Energy, or any other government personnel. This condition has been honored throughout the process and the writing team has worked with full independence.

The group began working in early April with a deadline of May 28. The short timeline and the technical nature of the material meant that we could not comprehensively review all topics. Rather, we chose to focus on topics that are treated by a serious, established academic literature; that are relevant to our charge; that are downplayed in, or absent from, recent assessment reports; and that are within our competence.

While the report is intended to be accessible to non-experts, we have omitted some introductory or explanatory material that can easily be accessed elsewhere. Nor have we attempted to survey entire literatures related to the topics covered. We have focused as much as possible on literature published since 2020 and referenced previous IPCC and NCA assessment reports. We have also used data through 2024 where possible.

The writing team is grateful to Secretary Wright for the opportunity to prepare this report and for his support of independent scientific assessment and open scientific debate.

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**PART I: DIRECT HUMAN INFLUENCE ON ECOSYSTEMS AND THE
CLIMATE**

1 CARBON DIOXIDE AS A POLLUTANT

Chapter summary:

Carbon dioxide (CO₂) differs in many ways from the so-called Criteria Air Pollutants. It does not affect local air quality and has no human toxicological implications at ambient levels. It is an issue of concern because of its potential effects on the global climate, a complex issue that occupies much of this report.

The Clean Air Act of 1970 defined six so-called Criteria Air Contaminants subject to regulation (EPA): particulate matter, ground-level ozone, sulfur dioxide, nitrogen dioxide, lead, and carbon monoxide. In 2007, the Supreme Court ruled that greenhouse gases (CO₂ among them) were also “pollutants” subject to regulation under Clean Air Act (*Mass. v. EPA*, 2007). While the definition of “pollutant” is ultimately a legal matter, there are important scientific distinctions between CO₂ and the Criteria Air Contaminants. The latter are subject to regulatory control because they cause local problems depending on concentrations that include nuisances (odor, visibility), damage to plants, and, at high enough exposure levels, toxicological effects in humans. In contrast, CO₂ is odorless, does not affect visibility and has no toxicological effects at ambient levels. It is a naturally occurring part of the atmosphere and a key component of human and plant respiration. CO₂ is essential for plant photosynthesis and higher levels are beneficial for vegetation. In these aspects, CO₂ is similar to water vapor.

Ambient outdoor air today contains about 430 parts per million (ppm) CO₂, increasing at about 2 ppm per year. The U.S. Occupational Safety and Health Administration issues guidelines for indoor workplaces in which elevated CO₂ might be encountered, such as where dry ice is used. The Permissible Exposure Limit is 5,000 ppm over 8 hours (OSHA, 2024). Allen *et al.* (2015) reported evidence of diminished performance on some cognitive tasks among workers in office cubicles when exposed to CO₂ levels above 1,000-1,500 ppm. These levels are far larger than any plausible ambient outdoor value through the end of the 22nd century.

The growing amount of CO₂ in the atmosphere directly influences the earth system by promoting plant growth (global greening), thereby enhancing agricultural yields, and by neutralizing ocean alkalinity. But the primary concern about CO₂ is its role as a greenhouse gas (GHG) that alters the earth’s energy balance. How the climate will respond to that influence is a complex question that will occupy much of this report.

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2 DIRECT IMPACTS OF CO₂ ON THE ENVIRONMENT

Chapter summary:

CO₂ enhances photosynthesis and improves plant water use efficiency, thereby promoting plant growth. Global greening due to increased CO₂ levels in the atmosphere is well-established on all continents.

CO₂ absorption in sea water makes the oceans less alkaline. The recent decline in pH is within the range of natural variability on millennial time scales. Most ocean life evolved when the oceans were mildly acidic. Decreasing pH might adversely affect corals, although the Australian Great Barrier Reef has shown considerable growth in recent years.

2.1 CO₂ and global greening

The growing CO₂ concentration in the atmosphere has the important positive effect of promoting plant growth by enhancing photosynthesis and improving water use efficiency. That is evident in the “global greening” phenomenon discussed below, as well as in the improving agricultural yields discussed in Chapter 10.

2.1.1 Measurement of global greening

“Greening” refers to an increase in the fraction of the Earth’s surface covered by plants. It can be quantified by the “Leaf Area Index” (LAI) measured by satellite. Many studies over the past decade have confirmed a global greening pattern (increase in LAI) attributable to rising CO₂ levels. Zhu *et al.* (2016) was one of the first studies to report that global greening was detectable using satellite sensors. From 1982 to 2011 they detected greening over 25-50 percent of the Earth versus “browning” over only four percent and attributed 70 percent of the greening to rising CO₂ levels (see Figure 2.1).

Zheng (2017) confirmed the pattern, noting that greening over thirty years had added 8 percent to global leaf area and that greening was mitigating warming. Greening has been observed globally but Chen *et al.* (2019) noted that China and India were exploiting it through land management changes. Thus, while China accounts for only 6.6 percent of global vegetated area it accounts for 25 percent of global net increase in LAI. Piao *et al.* (2020) noted that greening was even observable in the Arctic.

While plant models predict increased photosynthesis in response to rising CO₂, Haverd *et al.* (2020) reported a CO₂ fertilization rate nearly double model predictions. That is, CO₂ fertilization had driven an increase in observed global photosynthesis by 30 percent since 1900, versus 17 percent predicted by plant models. This indicates that global models of the socioeconomic impacts of rising CO₂ have likely understated the benefits to crops and agriculture. The connection between CO₂ fertilization and agriculture will be discussed in Chapter 9.

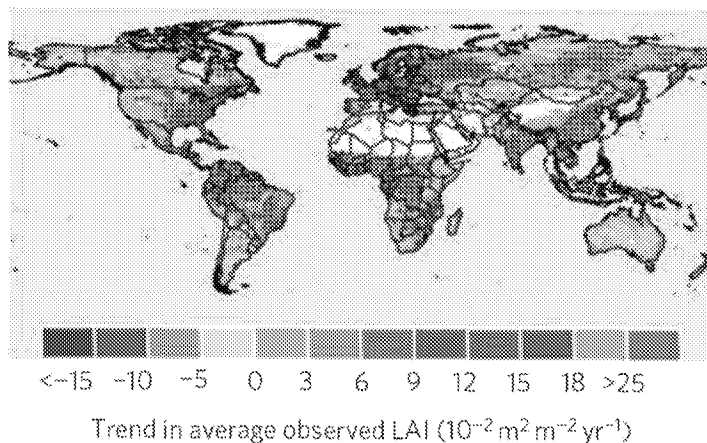


Figure 2.1: Trends in average Leaf Area Index (LAI). Source: Zhu *et al.* 2016 Figure 3.

Finally, Chen *et al.* (2024) surveyed claims that greening had slowed since 2000 due to drought stress. They found instead that faulty satellite sensors were the cause and that the corrected data showed accelerated greening over 55 percent of the globe while browning was accelerating over only 7 percent. In other words, the greening trend continued with no evidence of slowdown, and CO_2 fertilization remained the dominant driver.

2.1.2 Photosynthesis and CO_2 levels

Plants build biomass through *photosynthesis*, a process that converts carbon dioxide, water, and light into sugar. The plant enzyme responsible for photosynthesis is Ribulose-1,5-bisphosphate-carboxylase/oxygenase or “Rubisco”. Photosynthesis is initiated when CO_2 is available at the surface of the Rubisco enzyme where it is converted to a molecule with 3 carbon atoms and thereafter incorporated into plant mass. This is referred to as the “C3” process.

Rubisco is estimated to have evolved about 3 billion years ago. Over geological time the Earth’s atmospheric CO_2 levels were usually many times higher than they are today, as shown in Figure 2.2.

About 400 million years ago CO_2 levels were an estimated 2,000-4,000 ppm and were at or above 1,000 ppm for much of the interval from 200 to 50 million years ago (Berner 2006, Judd *et al.* 2024). Over the past 35 million years the level of atmospheric CO_2 has been steadily declining, falling to as low as 170 ppm during glaciations (Gerhart and Ward 2010).

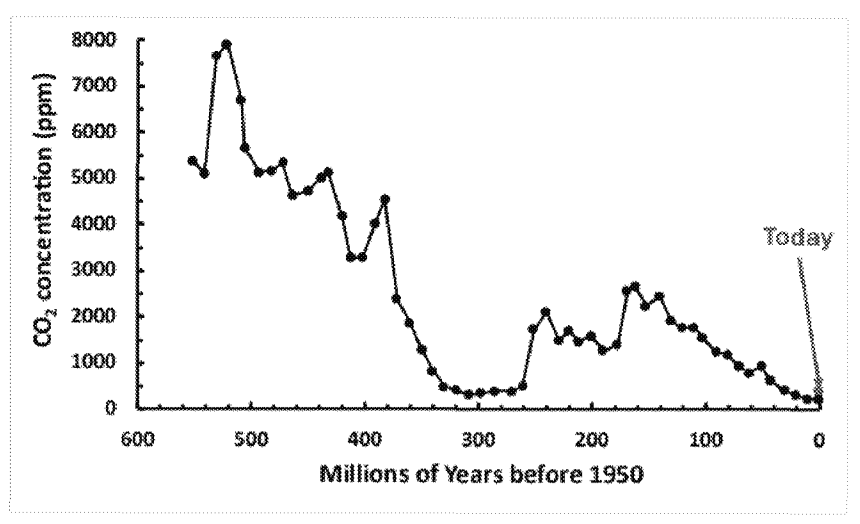


Figure 2.2. Ambient CO₂ levels over the past 550 million years. Today's concentration of 430 ppm is also shown. Source: (Berner 2006)

In response to low-CO₂ conditions some plants evolved another photosynthetic pathway called C4, in which CO₂ is concentrated in the vicinity of Rubisco, allowing for the C3 process to function more efficiently. For agricultural purposes the plant categories are:

- C3: rice, wheat, soybeans and most other crops
- C4: maize (corn), sugar cane, millet, sorghum

Had atmospheric CO₂ levels continued declining, plant growth would have declined and eventually ceased. Below 180 ppm, the growth rates of many C3 species are reduced 40-60 percent relative to 350 ppm (Gerhart and Ward 2010) and growth has stopped altogether under experimental conditions of 60—140 ppm CO₂. Some C4 plants are still able to grow at levels even as low as 10 ppm, albeit very slowly (Gerhart and Ward 2010).

Current CO₂ levels are about 430 ppm, up from 280 ppm in the early 1800s. The positive response of plants to extra CO₂ is illustrated in Figure 2.3, reproduced from Gerhart and Ward (2010). It shows the growth effect of CO₂ on Velvetleaf (*Abutilon theophrasti*) seedlings over 14 days under controlled conditions where only the CO₂ exposure is varied. The gains induced by increasing CO₂ from 150 ppm to 350 ppm continue under a further doubling to 700 ppm.

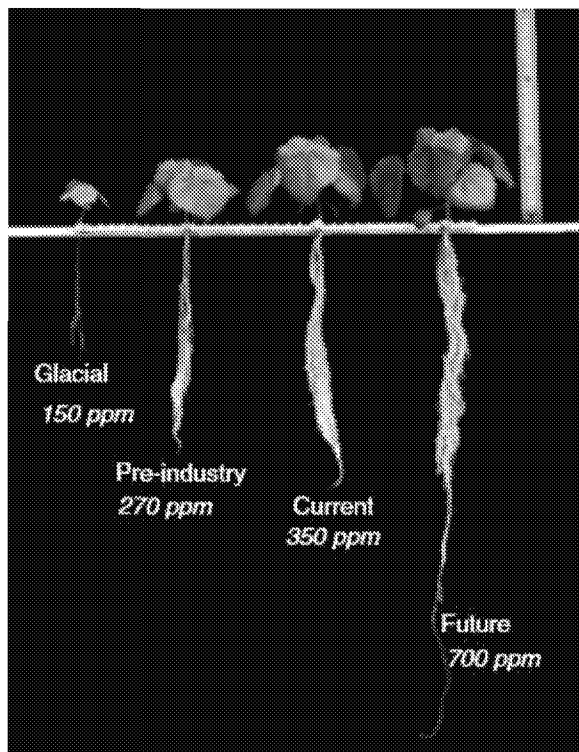


Figure 2.3: growth of *Abutilon theophrasti* after 14 days under identical conditions but for the indicated variations in CO₂ levels. Source: Gerhart and Ward (2010). Note “Current” corresponds to 1988 in image.

Over the past 60+ years there have been thousands of studies on the response of plants to rising CO₂ levels. The overwhelming theme is that plants, especially C3 plants, benefit from extra CO₂. There are two mechanisms by which CO₂ confers a growth benefit:

- Enhanced photosynthesis via the metabolic pathways described above.
- Increased water use efficiency. This arises because plants draw in CO₂ by opening the stomata (pores) on the leaf surface. When CO₂ is scarce the stomata must be kept wide open for long periods, allowing water to evaporate. Under enriched CO₂ conditions the stomata remain closed for longer periods, thus helping the plant retain water longer, and so increasing water use efficiency.

Specific effects of climate change on U.S. agriculture will be reviewed in Chapter 9.

2.1.3 Rising CO₂ and crop water use efficiency

Derying *et al.* (2016) surveyed evidence on crop water productivity (CWP), the yield per unit of water used, drawing attention to the potential for CO₂ both to enhance photosynthesis and to reduce leaf-level transpiration (water loss during leaf respiration). They surveyed all available FACE data (Free Air CO₂

Enrichment—see Chapter 9) on crop yield changes for maize (corn), wheat, rice, and soybean and combined it with crop model data simulating yield responses as of 2080 under the extreme RCP8.5 emissions scenario in four growing regions (Tropics, Arid, Temperate and Cold) each of which were split into rainfed and irrigated sub-regions. They reported that models without CO₂ fertilization predicted CWP losses in every region, but those were more than offset by CO₂ fertilization so that all regions showed a net CWP gain. Deryng *et al.* (2016) also reported that negative impacts of warming on wheat and soybean yields were fully offset by CWP gains and mitigated by up to 90 percent for rice and 60 percent for maize.

Similarly, Cheng *et al.* (2017) noted that increased global plant growth from 1982 to 2011 due to rising CO₂ uptake was accompanied by such large gains in CWP that global water use by plants had not increased, despite the extra biomass.

Deryng *et al.* (2016) assumed that climate change would “exacerbate water scarcity”. Yet while models do predict that drylands will expand under climate warming, current data show the opposite: greening is happening even in arid areas. Zhang *et al.* (2024) report that due to increased CO₂ levels “increasing aridity in drylands won’t lead to a general loss of vegetation productivity”; at most only 4 percent of currently arid areas will see increased desertification.

2.1.4 CO₂ fertilization benefits in IPCC Reports

The IPCC has only minimally discussed global greening and CO₂ fertilization of agricultural crops. The topic is briefly acknowledged in a few places in the body of the IPCC 6th and earlier Assessment Reports but is omitted in all Summary documents. Section 2.3.4.3.3 of the AR6 Working Group I report, entitled “global greening and browning,” points out that the IPCC Special Report on Climate Change and Land had concluded with *high confidence* that greening had increased globally over the past 2-3 decades. It then discusses that there are variations in the greening trend among data sets, concluding that while they have *high confidence* greening has occurred, they have *low confidence* in the magnitude of the trend. There are also brief mentions of CO₂ fertilization effects and improvements in water use efficiency in a few other chapters in the AR6 Working Groups I and II Reports.

Overall, however, the Policymaker Summaries, Technical Summaries, and Synthesis Reports do not discuss the topic.

2.2 The Alkaline Oceans

2.2.1 Changing pH

A neutral aqueous solution has a pH of 7.0, while one with pH greater than 7.0 is alkaline (also termed basic) and with pH less than 7.0 is acidic. The modern-day global average pH of sea water is estimated to be 8.04 (Copernicus Marine Service 2025), down from an estimated value of 8.2 in pre-Industrial times (Gattuso and Hansson, 2011). As CO₂ concentrations in the atmosphere increased, the oceans absorbed more, which decreases their pH. Depending upon the oceans’ buffering capacity, they are expected to become somewhat less alkaline over time, consistent with the observed decrease in pH.

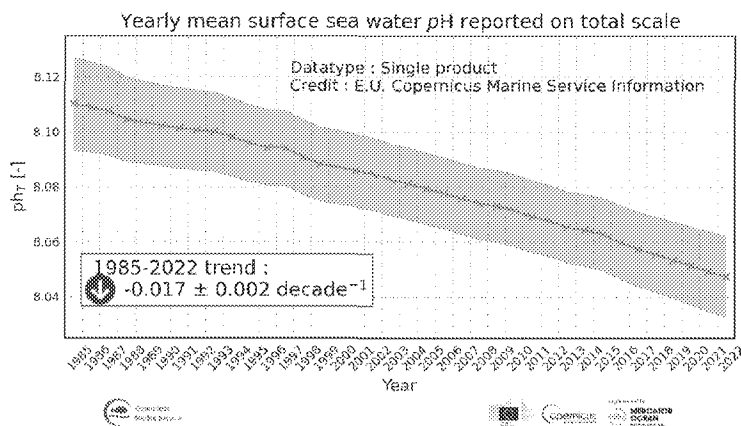


Figure 2.4: Ocean pH 1985 – 2022. Source: Copernicus Marine Service 2025

While this process is often called “ocean acidification”, it is a misnomer because the oceans are not expected to become acidic; “ocean neutralization” would be more accurate. Even if the water were to turn acidic, it is believed that life in the oceans evolved when the oceans were mildly acidic with pH 6.5 to 7.0 (Krissansen-Totton *et al.*, 2018). On the time scale of thousands of years, boron isotope proxy measurements show that ocean pH was around 7.4 or 7.5 during the last glaciation (up to about 20,000 years ago) increasing to present-day values as the world warmed during deglaciation (Rae *et al.*, 2018). Thus, ocean biota appear to be resilient to natural long-term changes in ocean pH since marine organisms were exposed to wide ranges in pH.

2.2.2 Coral reef changes

There are concerns that a decreasing pH of sea water will reduce the calcification rate of coral reefs. But coral reefs already endure large swings in pH, partly due to daily photosynthetic activity in the reef; measured pH values range from 9.4 during the day to 7.5 at night (Revelle and Fairbridge, 1957). De’ath *et al.* (2009) reported that a portion of Australia’s Great Barrier Reef (GBR, the world’s largest coral reef ecosystem) had experienced a 14 percent decline in calcification since 1990. This was tentatively attributed to increasing water temperature and decreasing pH. But Ridd *et al.* (2013) showed that report to have resulted from a biased data analysis that, when corrected, showed no change in calcification rates. Nevertheless, the alarm produced by the original paper has persisted as evidenced by the large number of published citations (541) to the original study compared to only 11 citations to the correction (as of 30 April 2025).

The most recent annual summary of GBR conditions from the Australian Institute of Marine Science indicates that coral production has rebounded strongly (AIMS, 2023). Figure 2.5 shows the results of the AIMS surveys of hard coral cover, expressed as a percentage of the reef area. Much of the decline in the GBR before 2011 turned out to be due to intense tropical cyclone activity (Beeden *et al.*, 2015) as well as a string of marine heatwaves, agricultural runoff and invasive species (Woods Hole, 2023). Given the reported declines in GBR calcification between 1990 and 2009 and the continued increase in atmospheric CO₂ levels the rebound has surprised some observers.

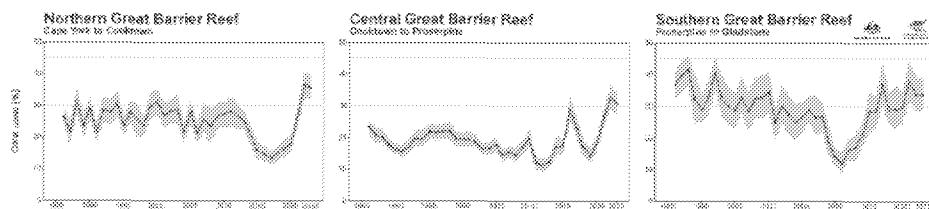


Figure 2.5 Hard coral cover of three regions of the Great Barrier Reef 1985 to 2023. Source: AIMS 2023.

It is being increasingly recognized that publication bias (alarming ocean acidification results preferred by high-impact research publications) exaggerates the reported impacts of declining ocean pH. An ICES Journal of Marine Science Special Issue addressed this problem with an article entitled, *Towards a Broader Perspective on Ocean Acidification Research*. In the Introduction to that Special Issue, H. I. Browman stated, “As is true across all of science, studies that report no effect of ocean acidification are typically more difficult to publish.” (Browman, 2016).

Similarly, a meta-analysis (Clements *et al.*, 2021) of the negative effects of ocean acidification on reef fish behavior found what they called a “decline effect”: initially dramatic conclusions published in prominent journals showing apparently large impacts of acidification tended to be followed up by subsequent studies on larger sample sizes yielding much smaller and typically non-existent effects. They call for their colleagues to improve research practices to counter the “decline effect”:

[The] vast majority of studies with large effect sizes in this field tend to be characterized by low sample sizes, yet are published in high-impact journals and have a disproportionate influence on the field in terms of citations. We contend that ocean acidification has a negligible direct impact on fish behavior, and we advocate for improved approaches to minimize the potential for a decline effect in future avenues of research (Clements *et al.*, 2021).

In summary, ocean life is complex and much of it evolved when the oceans were acidic relative to the present. The ancestors of modern coral first appeared about 245 million ago; as can be seen from Figure 2.2, CO₂ levels for more than 200 million years afterward were many times higher than they are today. Much of the public discussion of the effects of ocean “acidification” on marine biota has been one-sided and exaggerated. There is a known tendency for studies reporting dramatic effects on things like coral cover and fish behavior to get published in high-impact journals and receive widespread press coverage, then subsequent research rebutting those findings receives little attention.

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3 HUMAN INFLUENCES ON THE CLIMATE

Chapter Summary:

The global climate is naturally variable on all time scales. Anthropogenic CO₂ emissions add to that variability by changing the total radiative energy balance in the atmosphere.

The IPCC has downplayed the role of the sun in climate change but there are plausible solar irradiance reconstructions that imply it contributed to recent warming.

Climate projections are based on IPCC emission scenarios that have tended to exceed observed trends. Most academic climate impact studies in recent years are based upon the extreme RCP 8.5 scenario that is now considered implausible; its use as a business-as-usual scenario has been misleading.

Carbon cycle models connect annual emissions to growth in the atmospheric CO₂ stock. While models disagree over the rate of land and ocean CO₂ uptake, all agree that it has been increasing since 1959.

3.1 Components of radiative forcing and their history

3.1.1 Historical radiative forcing

A changing climate has been the norm throughout the Earth's 4.6-billion-year history. The Earth's temperature and weather patterns change naturally over timescales ranging from decades to millions of years. Natural variations in the surface climate originate in two ways. Internal climate fluctuations associated with circulations in the atmosphere and ocean exchange energy, water, and carbon between the atmosphere, oceans, land, and ice. External influences on the climate system include variations in the energy received from the sun and the effects of volcanic eruptions. Human activities influence climate through changing land use and land cover. Humans are also changing the composition of the atmosphere by emissions of CO₂ and other greenhouse gases and by altering the concentration of aerosol particles in the atmosphere.

The earth is warmed by the sunlight it absorbs and is cooled by the heat it radiates to space. Averaged over the Earth's surface, each of these processes involve power flows of about 240 Watts per square meter (W/m²). When they are in balance, there are no net external causes of warming or cooling. Both human and natural influences on the climate alter this balance and so cause the climate to change.

Influences on the Earth's energy balance at the top of the atmosphere are quantified by "radiative forcing", the extent to which they disturb the warming/cooling balance; a positive forcing warms while a negative forcing cools. The IPCC's estimated history of major components of radiative forcing since 1750 is shown in the following two figures from its AR6.

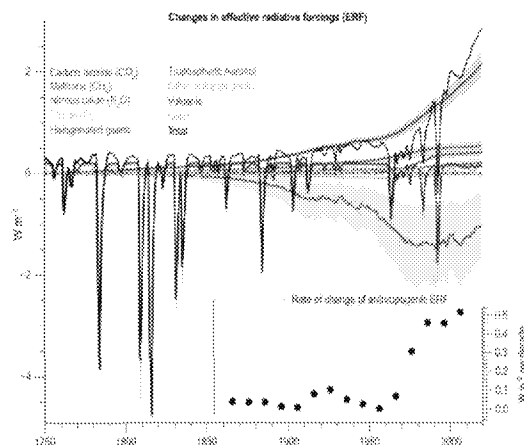


Figure 3.1.1: IPCC estimates of radiative forcing components over time. Shading indicates uncertainty ranges. Source: AR6 WGI Ch2 Fig. 10

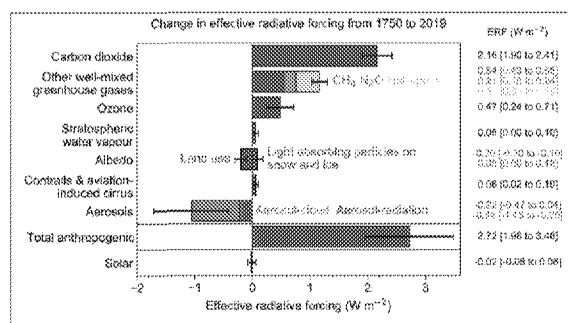


Figure 3.1.2 IPCC estimates of radiative forcing component changes from 1750 to 2019. Source: AR6 WGI Ch 7 Fig. 7-6.

These graphs show that the total radiative forcing is comprised of both natural and anthropogenic components. Carbon dioxide is the largest human influence on the climate and the one most relevant to the influence of fossil fuel use. It exerts a warming influence by decreasing the cooling power of the atmosphere. Emissions of CO₂ are accumulating in the atmosphere, as described in the following section, so that the warming influence is growing. Other greenhouse gases (methane, nitrous oxide, halogens, and ozone) act similarly, currently adding another 75 percent to CO₂'s warming. Aerosols exert an overall cooling effect, although with large uncertainties in the way they catalyze the formation of reflective clouds. As a result, understanding the causes of recent warming requires not just identifying the warming effects of CO₂, but also the more uncertain cooling effects of aerosols.

The IPCC assesses the change in the radiative forcing by the sun to be negligible, based on their preference for data reconstructions that imply minimal solar change since preindustrial times. But Connolly *et al.* (2021) reviewed sixteen different Total Solar Irradiance (TSI) reconstructions in the literature

covering the years 1600-2000; the reconstructions vary from almost no change in TSI to a relatively large upward trend. Those authors note that the variation in TSI reconstructions combined with variations in surface temperature reconstructions allows for inferences consistent with either no or most 20th century warming being attributable to the sun.

A particularly thorny issue is the gap in TSI data between 1989 and 1991 due to a delay in the launch of a monitor following the Space Shuttle Challenger disaster on January 28 1986. This delay prevented a replacement satellite from being launched in time to overlap with, and its readings to be intercalibrated with, the prior system (Zacharias 2014, Scafetta *et al.* 2019). This is called the ACRIM (Active Cavity Radiometer Irradiance Monitor) gap problem. The question of whether there is an upward trend in TSI over 1978 to 2018 hangs on how the ACRIM data gap is filled. Connolly *et al.* (2021) found that the IPCC's consensus statements on solar forcing were formulated prematurely through the suppression of dissenting scientific opinions.

Another natural radiative forcing component is volcanic aerosols, which exert an episodic cooling influence. Box 4.1 in the IPCC AR6 Report addresses the climate impact of volcanic eruptions, noting three explosive volcanic eruptions that occurred in the first half of the 19th century. This included the 1815 Tambora eruption that resulted in the 'year without summer', with multiple harvest failures across the Northern Hemisphere. There is uncertainty about the sign of the relatively small forcing due to the submarine volcano Hunga Tonga which erupted in 2022 (Jenkins *et al.* 2023, Schoeberl *et al.* 2024).

Figure 3.1.1 shows that the anthropogenic forcing component was negligible before about 1900 and has increased steadily since, rising to almost 3 W/m² today. However, this is still only about 1 percent of the unperturbed radiation flows, making it a challenge to isolate the effects of anthropogenic forcing; state-of-the-art satellite estimates of global radiative energy flows are only accurate to a few W/m².

Natural sources of global energy imbalance other than volcanoes and total solar irradiance (TSI) are not included in these graphs because they remain largely unknown.

3.1.2 Change in atmospheric CO₂ since 1958

Carbon dioxide's warming influence depends on how much "extra" CO₂ accumulates in the atmosphere- *i.e.*, its concentration above the preindustrial value of 280 ppm. The CO₂ level as recorded at the Mauna Loa observatory in Hawaii, generally used as the representative global average concentration, is available online at [HYPERLINK "<https://gml.noaa.gov/ccgg/trends/index.html>"]. The concentration was about 316 ppm at the start of the record in 1959 and is now about 430 ppm, a 36 percent increase. At the end of the last glaciation CO₂ levels had fallen to about 180 ppm. As discussed in Chapter 2, C3 plants begin dying at CO₂ levels below about 140 ppm and C4 plants at levels below 100 ppm, so if CO₂ levels had continued falling plant life would have been imperiled.

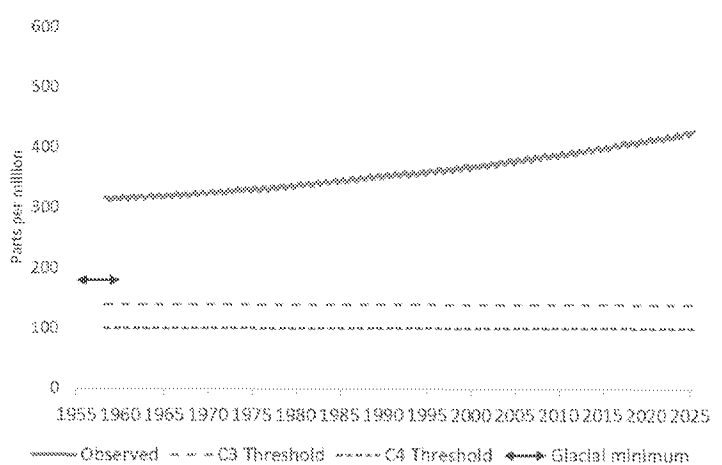


Fig. 3.1.3. Yearly average atmospheric CO₂ concentrations(1959-2025) in ppm measured at Mauna Loa (blue). C3 Threshold: Level below which C3 plants begin dying (140 ppm, see Chapter 2). C4 Threshold: Level below which C4 plants begin dying (100 ppm, see Chapter 2). Glacial minimum: Minimum level during recent glaciations (purple arrow). CO₂ data source: [[HYPERLINK "https://gml.noaa.gov/ccgg/trends/index.html"](https://gml.noaa.gov/ccgg/trends/index.html)]

The annual increase in concentration is only about half of the CO₂ emitted because land and ocean processes currently absorb “excess” CO₂ at a rate approximately 50 percent of the human emissions. Future concentrations, and hence future human influences on the climate, therefore depend upon two components: (1) future rates of global human CO₂ emissions, and (2) how fast the land and ocean remove extra CO₂ from the atmosphere. We discuss each of these in turn.

3.2 Future emission scenarios and the carbon cycle

3.2.1 Emission scenarios

Assessing the dangers of future GHG emissions requires assumptions about what those emissions will be. Future emissions, and hence human influences on the climate, will depend upon future demographics, economic activity, regulation, and energy and agricultural technologies. Various assumptions about each of those lead to projections of greenhouse gas emissions and concentrations, aerosol concentrations, and changes in land use, which ultimately can be combined into assumptions about anthropogenic radiative forcing.

The great uncertainties about these many factors make it impossible to precisely predict future emissions. Instead, the IPCC has created various sets of scenarios meant to span a plausible range of possibilities for population, economy, and technologies. Recent versions of the scenarios are labeled by a number indicating the anthropogenic radiative forcing expected in 2100 under that scenario. Thus, a scenario labeled with a “6” corresponds to 6 W/m² of human-induced radiative forcing (warming) at the end of the century. (Recall current anthropogenic radiative forcing is about 2.7 W/m².)

Although the IPCC does not claim its emission scenarios are forecasts, they are often treated as such. Comparisons of past scenario groups against observations show that IPCC emission projections tend to overstate actual subsequent emissions. For the IPCC Third and Fourth Assessment Reports a set of emission projections from the *Special Report on Emission Scenarios* was used; these were referred to as the SRES scenarios. McKittrick *et al.* (2012) showed that, when converted to per capita values, the SRES scenario emissions distribution was skewed upwards and incompatible with observed trends. McKittrick and Wood (2013) further showed that global energy market dynamics constrained global per capita CO₂ emissions to an extent not reflected in IPCC scenarios. The bias of the SRES scenarios was confirmed by the later analysis of Hausfather *et al.* (2019) who showed that observed atmospheric CO₂ concentrations tracked the low end of the SRES range and also of subsequent IPCC scenario ranges (Figure 3.2.1).

For AR5 the IPCC developed a new set of scenarios called the Representative Concentration Pathways (RCPs). These were identified by a number representing the increase in forcing and were thus called RCP2.6, RCP4.5, RCP6.0 and RCP8.5. RCP2.6 (implying an anthropogenic radiative forcing in 2100 of 2.6 W/m²) describes a GHG concentration pathway leading to warming well below 2°C. At the other end of the scale RCP8.5 is an extreme outcome implying nearly 5°C warming from 1900 to 2100.

RCP8.5 came to be referred to as the no-policy baseline, or “business-as-usual” scenario in both the academic literature and popular media. It was therefore used to generate the reference outcome supposedly representing the 21st century world in the absence of increasingly stringent emission reduction policies. But RCP8.5 was intended as a low-probability high emissions scenario and its use as a business-as-usual baseline has been criticized as grossly misleading.¹ Hausfather and Peters (2020) writing in a commentary in *Nature*, pointed out that RCP8.5 was developed as an extreme worst-case, and its misuse as a “business as usual” baseline has resulted in a large number of misleading studies and media reporting.

The implausibility of the RCP8.5 scenario was examined by Burgess *et al.* (2021). They noted that RCP8.5 has already diverged from observed trends in energy use and the near future trends diverge sharply from those of the International Energy Agency (IEA), which provides market-based projections of energy use for the coming decades. The IEA trends, which are based on plausible extrapolations of current economic and demographic trends, run below the entire envelope of RCP projections, implying not only that RCP8.5 is dubious, but that even lower emission scenarios are unlikely.

Widespread use of RCP8.5 has created a bias towards alarm in the climate impacts literature. The extent of this problem was confirmed in a literature analysis by Pielke Jr. and Ritchie (2020). They found that some 16,800 scientific papers published between 2010 and 2020 used the RCP8.5 scenario, with about 4,500 of the articles linking RCP8.5 to the concept of “business-as-usual”. Their analysis showed how RCP8.5 was misused not only by individual researchers, but also by influential science agencies like the IPCC and the U.S. National Climate Assessment (USNCA), which has directly led to misleading coverage in prominent media outlets.

¹ This extreme scenario is useful for modelers, since a large forcing generates a large response (warming) making it easier to assess a model’s sensitivity. But that’s very different than claiming it is a plausible future outcome.

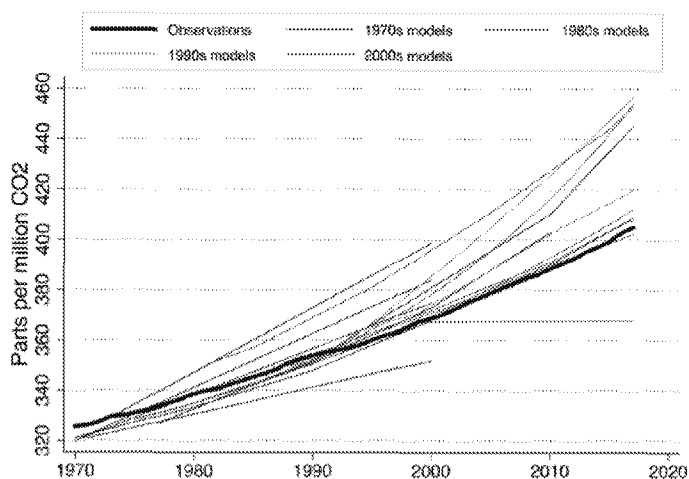


Figure 3.2.1 Comparison of observed atmospheric CO₂ levels (black line) with scenario projections (colored lines, mainly IPCC) since the 1970s. Source: Hausfather *et al.* (2019) Figure S4.

Pielke and Ritchie (2020) reported new studies using RCP8.5 were published at a rate of about 20 per day, with about two per day specifically linking RCP8.5 and “business as usual.” They conclude that the climate research community has spent a decade “committing scientific resources to science fiction” and that “The scientific literature has become imbalanced in an apocalyptic direction.”

The IPCC developed a new set of scenarios for AR6, the “Shared Socioeconomic Pathway” (SSP) scenarios, which have continued the bias shown in the RCP and SRES scenarios. Figure 3.2.2 shows total global observed CO₂ emissions compiled by the International Energy Agency (IEA) merged with the emission projection from the IEA taking account of energy use projections and current policies. The other lines show the range of IPCC SSP scenarios (SSP1-SSP5). As of 2023, global CO₂ emissions have been tracking well below SSP7.0 and are even below SSP2-4.5.

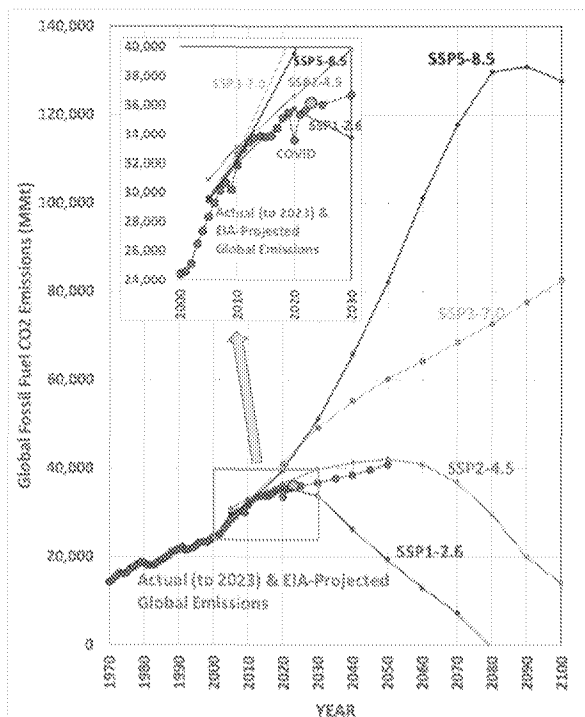


Figure 3.2.2. Observed and projected CO₂ emissions. Source: IPCC (SSP scenarios) and Energy Information Administration (EIA). Green: observed historical emissions and EIA projections. Other lines: SSP1-5. Data source: Friedlingstein *et al.* (2024).

3.2.2 The carbon cycle relating emissions and concentrations

Carbon dioxide emissions from fossil fuel burning (and to a lesser extent deforestation and cement production) have led to steadily increasing CO₂ concentrations in the atmosphere, as shown in Fig. 3.1.3. The relation between emissions and concentration is determined by the global carbon cycle of land and ocean processes that exchange carbon with the atmosphere. Our understanding of these processes was reviewed by Crisp *et al.* (2021).

There are about 850 Gt of carbon (GtC) in the Earth's atmosphere², almost all of it in the form of CO₂. Each year, biological processes (plant growth and decay) and physical processes (ocean absorption and outgassing) exchange about 200 GtC of that carbon with the Earth's surface (roughly 80 GtC with the land and 120 GtC with the oceans). Before human activities became significant, removals from the atmosphere were roughly in balance with additions. But burning fossil fuels (coal, oil, and gas) removes carbon from the ground and adds it to the annual exchange with the atmosphere. That addition (together with a much

² Because CO₂ is chemically transformed through the course of the carbon cycle, it is more convenient to track atoms of carbon rather than molecules of CO₂. One gigatonne of carbon (GtC) is equivalent to about 3.7 Gt of CO₂.

smaller contribution from cement manufacturing) amounted to 10.3 GtC in 2023, or only about 5 percent of the annual exchange with the atmosphere.

The carbon cycle accommodates about 50 percent of humanity's small annual injection of carbon into the air by naturally sequestering it through plant growth and oceanic uptake, while the remainder accumulates in the atmosphere (Ciais *et al.*, 2013). For that reason, the annual increase in atmospheric CO₂ concentration averages only about half of that naively expected from human emissions.

To project future CO₂ concentrations in the atmosphere, and hence future human influences on the climate, it is important to know how the carbon cycle might change in the future. The historical near constancy of that 50 percent fraction means that the more CO₂ humanity has produced, the faster nature removed it from the atmosphere. That 50 percent fraction changes somewhat from year to year due to natural carbon cycle imbalances from El Niño, La Niña, and varying weather patterns. There was also a substantial additional reduction in atmospheric CO₂ after the 1991 eruption of Mt. Pinatubo, a curious result that has yet to be explained (Angert *et al.*, 2004).

The main processes that remove excess CO₂ from the atmosphere are increased growth of land vegetation (especially at high latitudes), some increase in the sequestering of carbon in soils, and uptake of CO₂ by the ocean due to the increasing partial pressure of atmospheric CO₂ over that of CO₂ dissolved in the ocean. All twenty land carbon cycle models tracked by the Global Carbon Project (Friedlingstein *et al.*, 2024) show land processes have been removing excess CO₂ at an increasing rate since 1959. This is consistent with a “global greening” phenomenon (Chapter 2.1) observed by satellites since monitoring of global greenness began in 1982.

While land vegetation has been responding positively to more atmospheric CO₂, uptake of extra CO₂ by ocean biological processes remains too uncertain to be measured reliably. Our current understanding of these and many more carbon cycle processes was reviewed by Crisp *et al.* (2021).

CO₂ uptake by land processes

The uptake of extra CO₂ from the atmosphere by land surface processes (as also inferred from global greening) has been modeled with 20 different dynamic global vegetation models, the outputs of which are updated every year by the Global Carbon Project (Friedlingstein, 2024). As seen in Fig. 3.2.3, all of those models agree that vegetation and soils have been sequestering carbon from the atmosphere. But we also see that the long-term trends over 1959 to 2023 (65 years) vary widely between models, by nearly a factor of 7. This demonstrates that there remains considerable uncertainty in how fast land processes are removing CO₂ from the atmosphere, which in turn creates uncertainty in future atmospheric CO₂ concentrations, which then produce uncertainty in climate model simulations of future climate change.

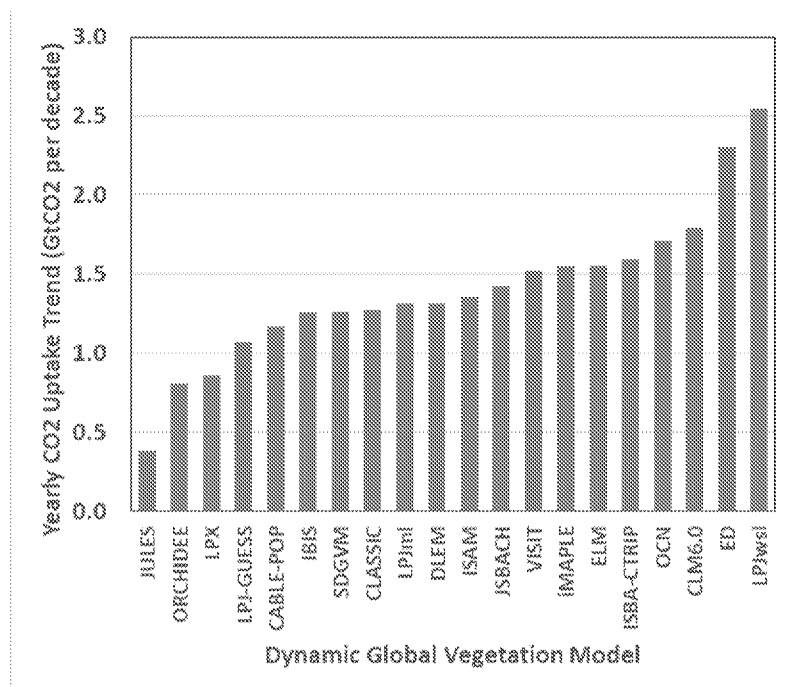


Figure 3.2.3 Trends of annual CO₂ uptake (GtCO₂ per year per decade) by land processes during 1959-2023 simulated by 20 different dynamic global vegetation models periodically reported by the Global Carbon Project (Friedlingstein, 2024).

CO₂ uptake by ocean processes

The uptake of extra CO₂ from the atmosphere by ocean processes has been modeled with 10 different ocean biogeochemistry models, the outputs of which are updated every year by the Global Carbon Project (Friedlingstein, 2024). Like the results from the land models, all of the ocean models agree that the global oceans have been sequestering carbon from the atmosphere at an increasing rate during 1959-2023 (Fig. 3.2.4). Unlike the land models, however, the ocean models show much better agreement with each other, with the model producing the fastest increasing CO₂ uptake being only 65 percent faster than the model with the most slowly increasing CO₂ uptake. In spite of the relative agreement among models, Friedlingstein *et al.* (2022) notes that there is substantial discrepancy between the different methods on the strength of the ocean sink over the last decade, particularly in the Southern Ocean.

Note that the average trend in CO₂ uptake across all land models in Fig. 3.2.3 is 25 percent larger than the average trend in ocean uptake. This suggests land processes are increasing in their ability to remove CO₂ faster than ocean processes are increasing their CO₂ sequestration.

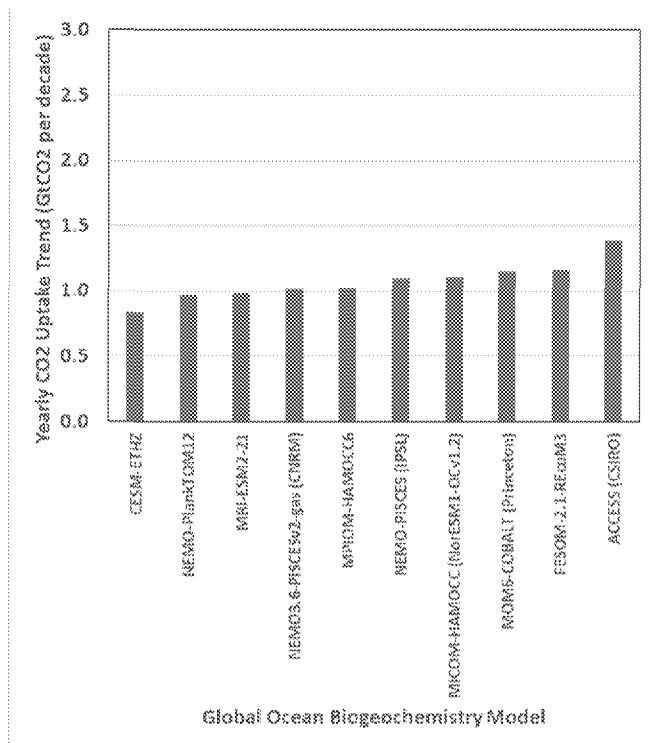


Figure 3.2.4 Trends of annual CO₂ uptake (GtCO₂ per year per decade) by ocean processes during 1959-2023 simulated by 10 different ocean biogeochemistry models periodically reported by the Global Carbon Project (Friedlingstein, 2024).

3.3 Urbanization influence on temperature trends

Historical temperature data over land has been collected mainly where people live. This raises the problem of how to filter out non-climatic warming signals due to Urban Heat Islands (UHI) and other changes to the land surface. If these are not removed the data might over-attribute observed warming to greenhouse gases. The IPCC acknowledges that raw temperature data are contaminated with UHI effects but claims to have data cleaning procedures that remove them. It is an open question whether those procedures are sufficient.

AR6 downplayed this issue by saying (WGI p. 235) that no recent evidence had emerged to alter the AR5 finding that urbanization causes an upward bias of no more than 10 percent in the global land surface warming trend. AR5 (WGI p. 189) also cited the 10 percent upper bound without citing a source. AR4 (WGI p. 244) cited Jones *et al.* (1990) and Peterson *et al.* (1999) as the basis of the claim. Peterson *et al.* had failed to find any difference in trends between rural and urban samples, although their definition of rural included local populations up to 10,000 persons while the relative influence of urbanization begins well below that (Spencer *et al.*, 2025). Jones *et al.* compared rural/urban warming in three regions: Eastern Australia, Eastern China and Western Soviet Union. Their definition of “rural” included towns of up to

10,000 in the former Soviet Union and up to 100,000 in China. They found relative warming biases greater than 10 percent in these areas but conjectured that the urbanization effect averaged over the areas they did not examine would bring the global land bias to under 10 percent of the observed warming trend.

Several papers appeared in print prior to the IPCC AR4 that argued that the warming effect of UHIs added a relatively large (30-50%) component to observed warming and was not simulated by climate models (de Laat and Maurellis 2006, McKittrick and Michaels 2007). These findings were based on correlations between locations of maximum warming over land with locations of maximum socioeconomic development. AR4 asserted (p. 244) that these correlations were an artefact of natural atmospheric circulations and were in fact statistically insignificant, and on that basis set the findings aside. Their claim was controversial because it was presented with no supporting evidence. McKittrick (2010) and McKittrick and Nierenberg (2010) showed that taking into account various conjectured alternative explanations for the correlations did not affect their significance. AR5 (p. 189) conceded that AR4 had provided “no explicit evidence” for its assessment and further acknowledged, based on these papers, that there was “significant evidence for such contamination of the record” i.e. a warming bias in the land record. However as already noted, elsewhere in the AR5 report they carried forward AR4 claim that it was less than 10 percent of observed warming. Further they provided no caution about using the land record for climate measurement despite conceding the evidence for UHI contamination. Recently Soon *et al.* (2023) estimated an urbanization bias in the Northern Hemisphere land record over 1850-2018 sufficient to increase the trend in the blended record from 0.55°C to 0.89°C per century.

Some studies providing evidence against UHI contamination compared warming rates between rural and urban locations (Jones *et al.* 1990, Peterson *et al.* 1999, Wickham *et al.* 2013). It is not known whether such methods would be capable of detecting UHI bias even when present. The influence of UHI warming is logarithmic in population, in other words it is strongest at low population density then levels off as local urbanization expands (Oke 1973, Spencer *et al.* 2025). Hence failure to find a difference in warming rates between urban and rural stations does not prove the absence of UHI contamination. McKittrick (2013) provided an empirical demonstration in which the rural/urban trends were not significantly different in a data set shown on other grounds to be contaminated with UHI bias.

Parker (2006) examined a sample of urban locations and found no difference in trends between subsets partitioned according to nighttime wind speed, concluding on this basis that urbanization could not be a significant factor. Here again the question is whether such a method would find UHI bias even if present. McKittrick (2013) presented an example in which UHI-contaminated data did not exhibit significant trend differences when stratified according to wind speed.

The challenge in measuring UHI bias is relating local temperature change to a corresponding change in population or urbanization, rather than to a static classification variable such as rural or urban. Spencer *et al.* (2025) used newly available historical population archives to undertake such an analysis and found evidence of significant UHI bias in U.S. summertime temperature data.

In summary, while there is clearly warming in the land record, there is also evidence that it is biased upward by patterns of urbanization and that these biases have not been completely removed by the data processing algorithms used to produce climate data sets.

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PART II: CLIMATE RESPONSE TO CO₂ EMISSIONS

4 CLIMATE SENSITIVITY TO CO₂ FORCING

Chapter Summary

There is growing recognition that climate models are not fit for the purpose of determining the Equilibrium Climate Sensitivity (ECS) of the climate to increasing CO₂. The IPCC has turned to data-driven approaches including historical data and paleoclimate reconstructions, but their reliability is diminished by data inadequacies.

Data-driven ECS estimates tend to be lower than climate model-generated values. The IPCC AR6 upper bound for the likely range of ECS is 4.0°C, lower than the AR5 value of 4.5°C. This lowering of the upper bound seems well justified by paleoclimatic data. The IPCC AR6 lower bound for the likely range of ECS is 2.5°C, substantially higher than the AR5 value of 1.5°C. This raising of the lower bound is less justified; evidence since the AR6 finds the lower bound of the *likely* range to be around 1.8°C.

4.1 Introduction

The climate's response to increasing concentrations of CO₂ is central to the scientific debate on anthropogenic climate change, and so also to the public debate on "climate action." The simplest measure of that response is the rise in the global average surface temperature, quantified by the Equilibrium Climate Sensitivity (ECS). ECS is defined as the amount of warming expected in response to a doubling of CO₂ from its pre-industrial concentration of 280 ppm, after all climate components have had time to adjust. Some components, like temperatures in the lower atmosphere (troposphere), adjust rapidly, while others such as the deep ocean and cryosphere might take as long as centuries. A related measure, the Transient Climate Response (TCR), better describes the shorter time scales; it is defined as the amount of warming when the CO₂ concentration is doubled by rising one percent annually for 70 years.

The 1979 Charney Report for the U.S. National Academy of Sciences (National Research Council 1979) proposed that the most likely ECS was $3.0 \pm 1.5^\circ\text{C}$. The IPCC repeatedly reaffirmed that range with only minor variations until its most recent AR6. AR5 termed 1.5–4.5°C as the *likely* range (66 percent probability) and stated that ECS is *extremely unlikely* (95 percent probability) to be below 1.0°C and *very unlikely* (90 percent probability) to exceed 6.5°C.

The uncertainty in ECS has remained stubbornly wide, despite many individual studies that claimed to narrow it (Hausfather 2023). Most recently, AR6 narrowed the *likely* range to 2.5–4.0°C and deemed the *very likely* range to be 2.0–5.0°C. This narrowing on the low end is disputed, as will be discussed below.

Uncertainties in ECS are highly consequential for policy making. As will be discussed in Chapter 11, economic models use ECS values to project the costs of CO₂ emissions. The traditional value (3.0 °C) has typically yielded modest global social costs of CO₂ emissions, sufficient to justify some policy actions, but mostly deferred to later in this century. If ECS is very high (above 4.5°C) immediate aggressive emission controls become more imperative, whereas no CO₂ emission controls are economically justifiable for ECS below 2.0°C (Dayaratna *et al.* 2016, 2020). Obtaining a precise estimate is impossible, so policy making needs to account for the uncertainty.

By itself, the equilibrium warming effect of a doubling of atmospheric CO₂ is slightly more than 1°C (Soden and Held 2006). Larger values of ECS arise from positive feedbacks that amplify the CO₂ warming. Water vapor feedback is positive: a warmer atmosphere might have more water vapor, which itself is a powerful greenhouse gas. Warmer temperatures also result in less snow and sea ice cover, allowing the

Earth to absorb more of the sun's radiation. Some simple estimates of these feedbacks increase the ECS to around 2°C (Sherwood *et al.*, 2020). Larger values of ECS are associated with positive cloud feedbacks.

Climate scientists use multiple lines of evidence to determine the Equilibrium Climate Sensitivity:

- Climate model simulations
- Historical observations
- Paleoclimatic reconstructions
- Process understanding of feedbacks

4.2 Model-based estimates of climate sensitivity

The ECS ranges given in IPCC AR4 and AR5 were obtained primarily by examining the behavior of large-scale climate models, also called General Circulation Models (GCMs). However, the IPCC changed course in its AR6 when it turned to a more direct data-driven methodology. Here we discuss some of the pitfalls of using GCMs to try to determine the Earth's climate sensitivity.

In principle, ECS is an emergent property of GCMs—that is, it is not directly parameterized or tuned, but rather emerges in the results of the simulation. Otherwise plausible GCMs and parameter selections have been discarded because of perceived conflict with ~~this an expected~~ warming rate, or aversion to a model's climate sensitivity being outside an accepted range (Mauritsen *et al.* 2012). This practice was commonplace for the models used in AR4; modelers have moved away from this practice with time. However, even in a CMIP6 model, Mauritsen and Roeckner (2020) state the following regarding their Max Planck Institute (MPI) climate model (emphasis added):

“We have documented how we tuned the MPI-ESM1.2 global climate model to match the instrumental record of warming; an endeavor which has clearly been successful. Due to the historical order of events, the choice was to do this practically by **targeting an ECS of about 3 K using cloud feedbacks**, as opposed to tuning the aerosol forcing.”

In other words, the MPI modelers chose an ECS value of 3°C and then tuned the cloud parameterizations to match their intended result.

As noted, direct warming from CO₂ doubling is only about 1°C (Soden and Held 2006); further warming arises from climate feedbacks that are not explicitly resolved by the GCM but rely on parameterizations of physical processes. Higher values of ECS arise primarily from assumed positive cloud feedbacks, whereas the magnitude and even the sign of the feedbacks are very uncertain. Elements of cloud feedback include changes in the latitudinal distribution of clouds, changes in the distribution of cloud height (changes in low versus high clouds), changes to the phase of clouds (ice versus liquid), changes in cloud particle size (associated with changes in concentration and/or composition of aerosol particles), changes in the precipitation efficiency of clouds, and even changes in how clouds are distributed over the daily solar cycle (Curry and Webster, 1999). It is difficult for GCMs to simulate any of these processes correctly owing to their small scale, let alone predict how they will change in the future. Further, cloud processes modulate the magnitudes of the water vapor, lapse rate, and the surface albedo feedbacks.

ECS can be determined from climate model simulations by doubling the concentration of CO₂ and allowing several centuries for the warming to equilibrate. To avoid the need for such long simulations, “effective climate sensitivity” is commonly evaluated from a 150-year simulation in response to a sudden quadrupling of CO₂.

The spread of ECS values from the CMIP5 ensemble of climate models used in AR5 was 2.0–4.7°C; that range increased for the CMIP6 models used in AR6 to between 1.8 and 5.7°C (Chen *et al.*, 2021, Scafetta 2021, see Figure 4.1). Far from settling-resolving the model-based climate sensitivity the range appears to be growing. The main cause of the overall upward shift in ECS in CMIP6 relative to CMIP5 is a larger positive cloud feedback, driven by changes to the cloud parameterizations in many CMIP6 models (Zelinka *et al.*, 2020)

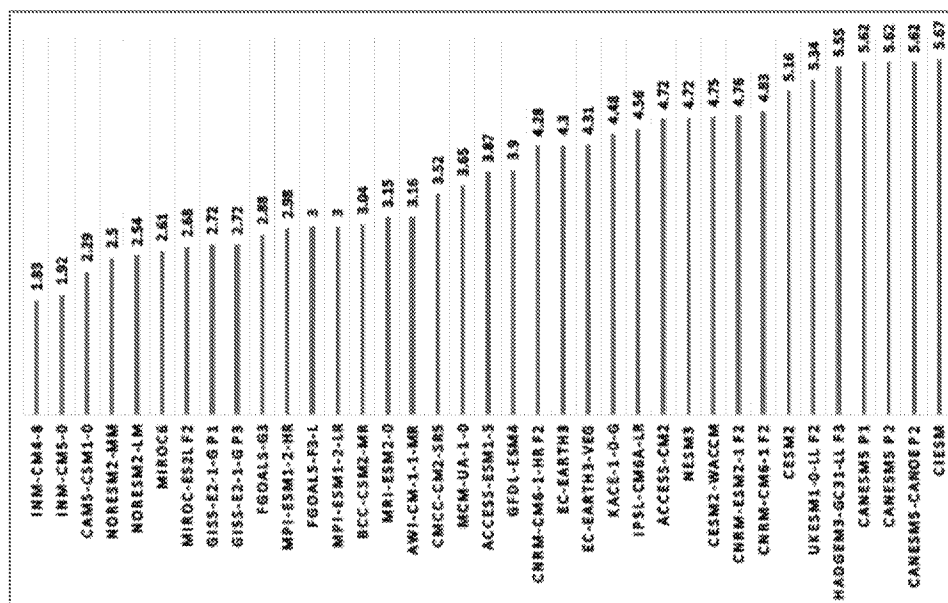


Figure 4.1 Equilibrium Climate Sensivities in °C of 37 climate models from the CMIP6 ensemble. Identifiers for the various models appear along the horizontal axis. From (Scafetta, 2021)

Because of concerns about model tuning and the high sensitivity to cloud parameterizations, the IPCC AR6 (2021) did not rely on climate model simulations in their assessment of climate sensitivity, relying instead on data-driven methods.

4.3 Data-driven estimates of climate sensitivity

Climate sensitivity can also be estimated from instrumental records of surface temperatures and ocean heat content, combined with estimates of how climate forcings (e.g., greenhouse gases, solar, volcanoes, aerosols) have changed in the past (Otto *et al.*, 2013). Using this information, a simple empirical model called an Energy Balance Model can be employed. It requires estimating a feedback parameter whose uncertainties are highly amplified in the resulting ECS (Roe and Baker, 2007).

The accuracy of the data-driven methods depends on the quality of the input data. Assumptions are needed about ocean heat storage, and good data is only available for recent decades. The greatest source of uncertainty is the amount and composition of aerosol particles and their interactions with cloud radiative properties (the so-called aerosol indirect effect; see Figures 3.1.1, 3.1.2). Climate models exhibit warming in response to GHGs but cooling in response to aerosols (Schwartz *et al.*, 2007). Observed 20th century warming can be shown to be consistent either with low ECS and low aerosol cooling, or high ECS and high aerosol cooling. Since fossil fuel use adds both GHGs and aerosols to the atmosphere, both effects need to be estimated to isolate the warming effect of CO₂.

Paleoclimate proxies are also used to evaluate the sensitivity of past climates by comparing paleoclimate changes in the Earth's temperatures to estimates of changes in forcings. The two most informative periods are the last glacial maximum (around 20,000 years ago), which was about 3–7°C colder than today, and a mid-Pliocene period (roughly three million years ago), which was 1–3°C warmer than today. The limits on cooling during the last glacial maximum give the best single evidence that high values of climate sensitivity are unlikely. However, paleoclimate estimates are associated with very large uncertainties in the estimated temperatures and forcings. Further, estimates of climate sensitivity based on past climate states might not be applicable to the current state of the climate system.

A recurring theme in the climate literature is that ECS estimates based on historical data are smaller than ECS estimates inferred from climate models (Sherwood and Forest 2024). About 15 estimates based on historical data appeared in the peer-reviewed literature between 2012 and 2024 yielding ECS best estimates between 1.0°C and 2.5°C, although critics have questioned some of the methods and the data quality. For AR6, the IPCC placed primary weight on the results of Sherwood *et al.* (2020) that combined historical data and paleoclimate proxies with the process-based approach and yielded a best estimate of 3.1°C with a *likely* range of 2.6–3.9°C. Lewis (2022) raised a number of concerns about this result, including methodological errors, outdated input values, and use of subjective Bayesian priors in the analysis. Lewis' analysis found that climate sensitivity is estimated to be much lower and better constrained than in the Sherwood *et al.* analysis – median 2.2°C (1.8–2.7°C in the 17–83 percent *likely* range, and 1.6–3.2°C in the 5–95 percent *very likely* range). The IPCC AR6 estimated only a 5 percent probability that ECS was below 2.3°C, whereas Lewis estimated it to be over 50 percent. The most recent publications on the debate between Sherwood *et al.* and Lewis further defend their respective positions: Sherwood and Foster (2024) and Lewis (2025).

An argument emphasized in AR6 is that data-driven ECS estimates might understate the future warming response to GHGs because of a so-called “pattern effect” (Forster *et al.*, 2021). The tropical Pacific is believed to strongly influence the overall efficiency with which the Earth radiates heat to space, but some regions remove heat more efficiently than others. If the west-to-east temperature gradient in the tropical Pacific is weakened in a warming climate, warming would concentrate where heat is less efficiently removed, raising ECS.

Most climate models simulate that rising GHGs will weaken the west-east temperature gradient, which led the IPCC in AR6 to conclude that data-driven ECS estimates understated the *future* ECS value. However, Seager *et al.* (2019) pointed out that, contrary to models, the west-east temperature gradient has been strengthening over time. They further argued that the mechanism predicting otherwise in climate models was based on a faulty characterization of oceanic dynamics and there is no reason to expect the gradient to weaken. A similar argument was recently made by Lee *et al.* (2024), who concluded that “the trajectory of the observed trend reflects the response to increasing GHG loading in the atmosphere”; in other words, GHG warming should lead to a future strengthening rather than a weakening of the temperature gradient. Increased efficiency of atmospheric cooling implies, if anything, that the future ECS in a warming climate might be *lower* than current estimates.

4.4 Transient Climate Response

The Transient Climate Response (TCR) provides a more useful observational constraint on climate sensitivity. TCR is the global temperature increase that results when CO₂ is increased at an annual rate of 1 percent over a period of 70 years (*i.e.*, doubled gradually). Relative to the ECS, observationally determined values of TCR avoid the problems of uncertainties in ocean heat uptake and the fuzzy boundary in defining equilibrium arising from a range of timescales for the longer-term feedback processes (*e.g.*, ice sheets). TCR is more generally related to peak warming and better constrained by historical warming, than ECS. IPCC AR6 judged the *very likely* range of TCR to be 1.2–2.4°C. In contrast to ECS, the upper bound of TCR is more tightly constrained. For comparison, the TCR values determined by Lewis (2023) are 1.25 to 2.0°C, showing much better agreement with AR6 values than was seen in a comparison of the ECS values.

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5 MODELS VERSUS OBSERVATIONS IN THE RECENT PAST

Chapter summary

Climate models show warming biases in many aspects of their reproduction of the past several decades. In response to estimated changes in forcing they produce too much warming at the surface (except in the models with lowest ECS), too much warming in the lower-and mid-troposphere and too much amplification of warming aloft.

Climate models also produce too much recent stratospheric cooling, invalid hemispheric albedos, too much snow loss, and too much warming in the Corn Belt. The IPCC has acknowledged some of these issues but not all.

5.1 Introduction

Climate models are the primary tool used to project future climate changes in response to increasing atmospheric levels of anthropogenic greenhouse gases. To assess the fitness of climate models for this purpose, it is reasonable to ask how well they reproduce the current climate and its variations over the past century. The box “Climate modeling” gives some detail on how climate models work.

Of great concern is the fact that, after several decades of the climate modeling enterprise involving approximately three dozen models operated by research centers around the world, the range of future warming they produce in response to a hypothetical doubling of atmospheric CO₂ extends over a factor of three, as we discussed in the previous chapter. This range of disagreement among models has not decreased for decades.

Problems with climate models are not just in their disagreement over the future, but also in their ability to replicate the recent past. Here we review some of the most important metrics of climate model accuracy: ability to reproduce historical surface, tropospheric and stratospheric temperature trends; ability to reproduce the vertical warming profile; and ability to reproduce other climate features such as snowfall. In all cases a persistent finding is that models on average err on the side of too much warming in response to estimated historical forcings.

BOX: Climate modeling

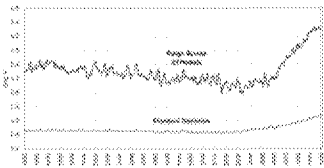
Most climate models, except for the simplest ones, represent Earth's land surface using a grid of squares on the order of 100 km wide. To simulate the atmosphere, typically 30 or more grid boxes are stacked above these squares. The ocean is modeled using a similar but finer grid, resulting in tens of millions of grid boxes for the atmosphere and oceans.

The computer models, based on physical laws, calculate how air, water, and energy move between grid boxes over time. The time step can be as small as 10 minutes, and repeating this process millions of times allows the simulation of climate over centuries. Running these models, even on the most powerful supercomputers, can take months. Comparing simulation results with historical climate data helps assess the accuracy of a model, while projections into the future estimate climate changes under assumed human and natural influences.

Despite sounding straightforward, climate modeling is highly complex. Many critical processes occur at scales smaller than the grid size. For instance, the flows of sunlight and heat in the atmosphere depend strongly on cloud cover. Since tracking individual clouds is impractical, researchers must make "subgrid" assumptions about the distribution of clouds in each grid box. Snow and ice cover, which affects how much sunlight is reflected or absorbed by the surface, is another subgrid factor.

Each subgrid assumption requires numerical parameters, which must be carefully set. Modelers initially estimate these parameters based on physics and observed climate patterns, then run the model. Because early results often diverge significantly from real-world observations, they "tune" these parameters to better match observed climate features. Different modeling teams use distinct assumptions and tuning strategies, leading to varied outcomes. Tuning is a necessary but delicate aspect of climate modeling, as it is for any complex system. Poor tuning can result in inaccurate simulations, while excessive tuning risks artificially steering results toward predetermined conclusions.

The spread of model indicators—Earth's 1880 (Figure 5.1), warming was only processes.



very wide. One of the most basic about 3°C across CMIP6 models prior to to over 4 °C. For comparison 20th century tantial differences among models' physical

Figure 5.1 CMIP6 Average surface temperature range across 33 models and standard deviation using SSP5-85 scenario. Data from KNMI Climate Explorer website at [[HYPERLINK "https://climexp.knmi.nl/start.cgi"](https://climexp.knmi.nl/start.cgi) \t "_blank"].

Beyond the models' ability to reproduce features of today's climate, the critical issue for society is how well they predict responses to subtle human influences, such as greenhouse gas emissions, aerosol cooling, and land-use changes. The most crucial aspect that models must capture correctly is "feedbacks." These occur when climate changes either amplify or suppress further warming. In general, the modeled net effect of all feedbacks doubles or triples the direct warming impact of CO₂.

5.2 Surface warming

A straightforward test of a climate model's validity is its ability to reproduce historical warming in response to known past changes in climate drivers such as greenhouse gases. Figure 5.2 is reproduced from Scafetta (2023), that groups the latest-generation (CMIP6) climate models into low ECS (1.5 to 3.0 degrees C), medium ECS (3.0 to 4.5 degrees C) and high ECS (4.5 to 6.0 degrees C), and compares their post-1980 global average temperature simulation ranges to those of three surface temperature records and one satellite-based lower troposphere temperature data product.

The leftmost column shows that the low-ECS models track the post-1980 historical warming record reasonably well, but the middle and right columns show that the medium- and high-ECS models conspicuously over-predict the warming.

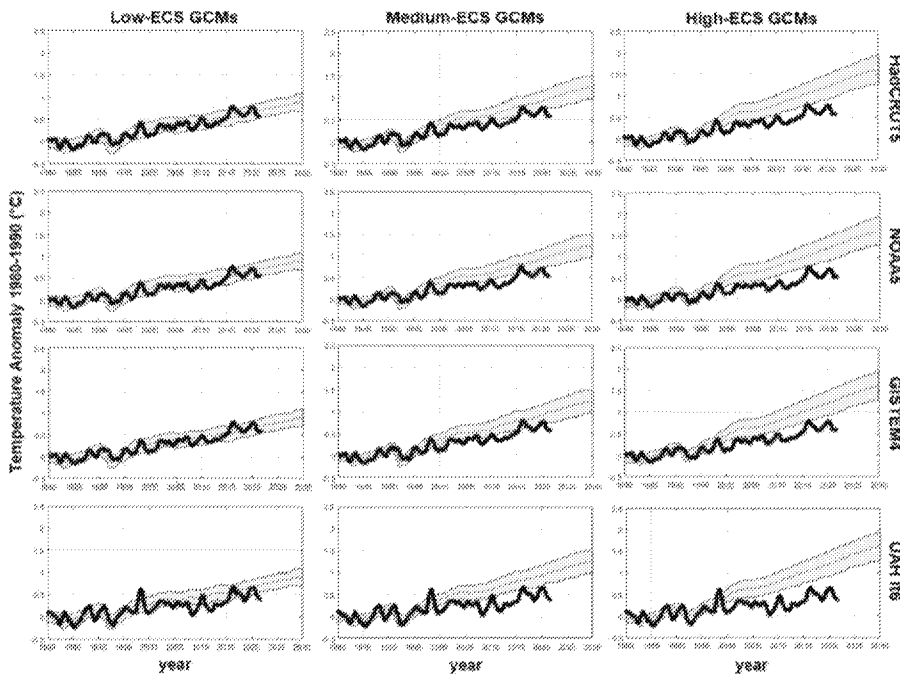


Figure 5.2: Model-Observational comparisons for Earth's surface warming. The columns correspond to model groups showing low-ECS to high-ECS, while the rows correspond to widely-used observed temperature records, the first three showing surface averages and the fourth showing the lower troposphere average. In each panel the yellow area denotes the mean and range ($\pm$ one standard deviation) of climate model simulations for that group. The thick black line shows the observed annual average temperature in the indicated source. Source: Scafetta (2023) Fig.2.

Spencer (2024) has also provided a useful summary of the Model-Observation mismatch by comparing trends in surface temperature data products with those in individual climate models, as summarized in Figure 5.3; most climate models show substantially more warming than the observations since 1979.

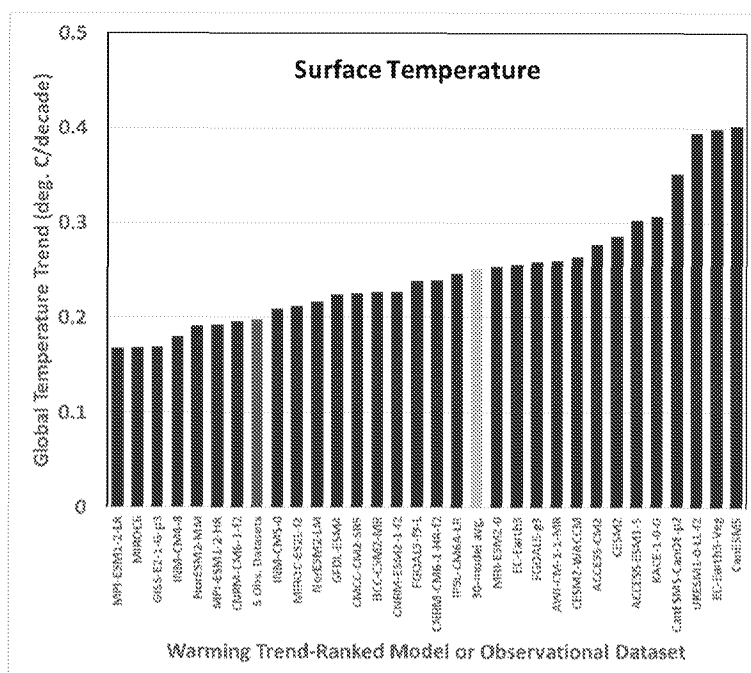


Figure 5.3 Global surface air temperature trends ($^{\circ}\text{C}/\text{decade}$), 1979–2024, from various CMIP6 climate models (red, 30-model average in orange); and the average of three thermometer datasets (HadCRUT5, NOAA Global Temp, and Berkeley 1 deg.) and two reanalysis datasets (ERA5 and NCEP/NCAR R1) in blue. Data source: <https://climexp.knmi.nl/start.cgi>.

5.3 Tropospheric warming

It has long been known that climate models on average overstate warming in the tropical troposphere. This region is an important test of climate models since this is where the signal of anthropogenic greenhouse warming emerges first and most strongly. Biases in tropospheric trends indicate model flaws in heat transfer processes that carry over to surface warming biases.

The discrepancy was flagged as a serious inconsistency in the first U.S. Climate Change Science Program report (Karl *et al.* 2006) and has been mentioned in every IPCC report since, but the discrepancy has gotten worse over time, and the bias is now global. McKittrick and Christy (2020) compared tropospheric warming trends in CMIP6 climate models to observed trends from satellites, weather balloons and reanalysis systems. Every model overpredicted the average observed warming trend over 1979-2014 in both Lower- and Mid-Troposphere layers, both globally and in the tropics. In most individual models the bias was statistically significant and on average across models it was highly significant.

Figure 5.4 presents the comparisons with data updated to 2024 (McKittrick and Christy 2025). The recent warm years moved the observed trend up slightly and widened the trend confidence intervals but the overall pattern remains the same: model bias is towards too much warming, in most cases the difference is statistically significant and on average the bias is statistically highly significant. McKittrick and Christy (2020) also showed that the bias is larger in high-ECS models, but even the models with lower average ECS predict too much warming. If future climate models were to realistically represent global tropospheric warming, they would likely be less sensitive than even the low-ECS members of the CMIP6 ensemble.

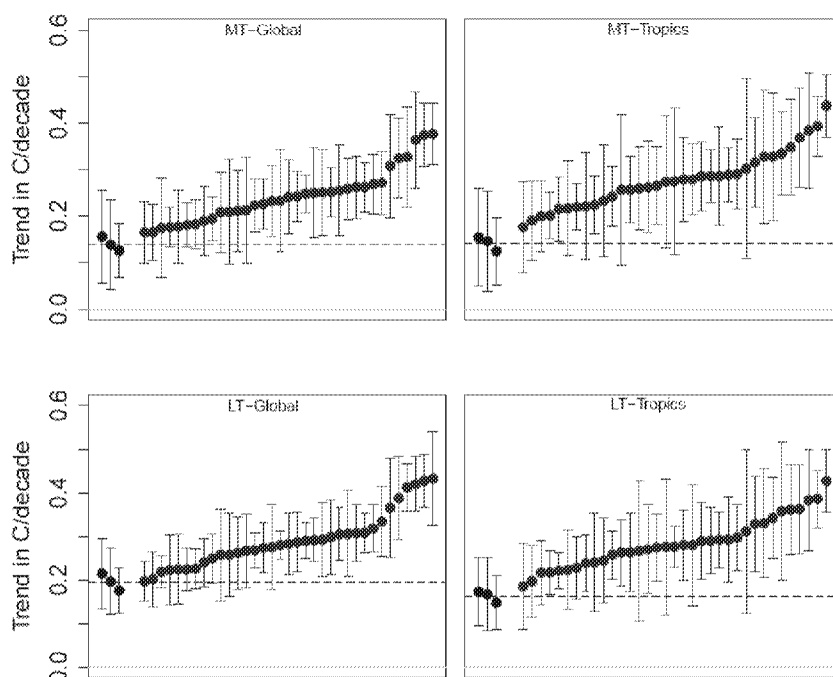


Figure 5.4: Observed versus CMIP6 modeled warming trends (°C/decade 1979-2024) in the global and tropical lower (LT) and mid-troposphere (MT). Uses methodology of McKittrick and Christy (2020) on data updated from 2014 to 2024. Blue dots: warming trends with 95 percent confidence intervals for 3 data products (radiosondes, reanalysis, and satellites). Blue dashed line: warming trend average for 3 observed series. Red dots: modeled warming trends with 95 percent confidence intervals in 35 models arranged lowest to highest.

As mentioned previously, the IPCC has long acknowledged the model-observation mismatch. For example, AR6 pp. 443-444 offers this on the tropical troposphere (it does not address the global comparison):

Several studies since AR5 have continued to demonstrate an inconsistency between simulated and observed temperature trends in the tropical troposphere, with models simulating more warming than observations (Mitchell *et al.*, 2013, 2020; Santer *et al.*, 2017a, b; McKittrick and Christy, 2018; Po-Chedley *et al.*, 2021)... Over the 1979–2014 period, models are more consistent with observations in the lower troposphere, and least consistent in the upper troposphere around 200 hPa, where biases exceed 0.1°C per decade. Several studies using CMIP6 models suggest that differences in climate sensitivity may be an important factor contributing to the discrepancy between the simulated and observed tropospheric temperature trends (McKittrick and Christy, 2020; Po-Chedley *et al.*, 2021), though it is difficult to deconvolve the influence of climate sensitivity, changes in aerosol forcing and internal variability in contributing to tropospheric warming biases (Po-Chedley *et al.*, 2021). Another study found that the absence of a hypothesized negative tropical cloud feedback could explain half of the upper troposphere warming bias in one model (Mauritsen and Stevens, 2015).

... In summary, studies continue to find that CMIP5 and CMIP6 model simulations warm more than observations in the tropical mid- and upper-troposphere over the 1979–2014 period (Mitchell *et al.*, 2013, 2020; Santer *et al.*, 2017a, b; Suárez-Gutiérrez *et al.*, 2017; McKittrick and Christy, 2018), and that overestimated surface warming is partially responsible (Mitchell *et al.*, 2013; Po-Chedley *et al.*, 2021). ... Hence, we assess with *medium confidence* that CMIP5 and CMIP6 models continue to overestimate observed warming in the upper tropical troposphere over the 1979–2014 period by at least 0.1°C per decade,

Notably, despite the accumulation of evidence of excess model warming the IPCC assigns only *medium confidence* to the existence of a warming bias.

5.4 Vertical temperature profile mismatch

Another important model-observational discrepancy is the excess amplification with altitude found in climate models. The comparison was in the IPCC AR5 Chapter 10, although only in the online supplement (Figure 10.SM.1) and only in a figure whose formatting obscured the point. Figure 10.SM.1 is not referenced in the main IPCC report nor in any summary so readers would not have been aware of it. Although not apparent at first glance, it shows that the 1979–2010 warming in the lower troposphere is so small as to be consistent with no GHG forcing at all and is inconsistent with the model runs that do have GHG forcing. In Figure 5.6 we adapt IPCC AR5 Figure 10.SM.1 to draw out this critical point.

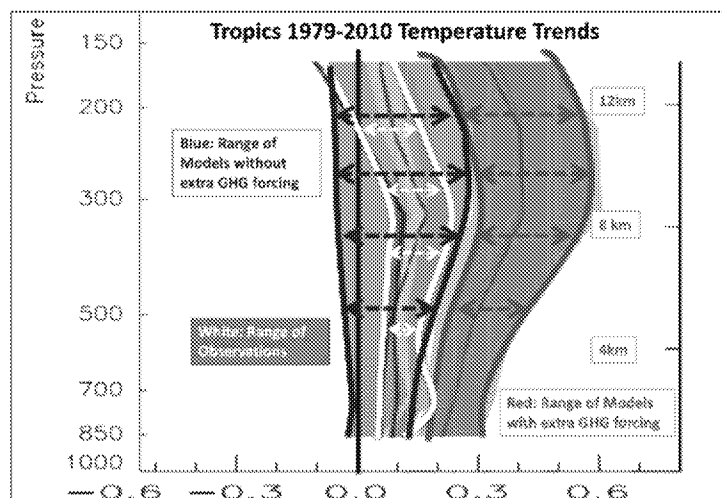


Figure 5.6: Vertical warming pattern for tropics (20S to 20N). Source: Annotated version of IPCC AR5 Figure 10.SM.1

Figure 5.6 compares model and observational temperature trends by altitude between 20S and 20N (the tropics). In this region where the models say the warming should be strongest, the observations (shown here in white) lie within the blue “No CO₂” band and entirely outside the “with CO₂” red envelope. This means that in the entire tropical atmospheric column from the surface to the base of the stratosphere, observed warming trends are so small as to be consistent with the output of models that have no anthropogenic CO₂, and inconsistent with the entire envelope of warming trends generated by models forced with increased CO₂.

A similar comparison is shown in Christy and McNider (2017), an updated version of which (covering 1979-2024) is reproduced as Figure 5.7. Modeled temperature trends exceed observations from the surface through the top of the troposphere, with observed trends below the entire model range at most pressure levels. Also shown in Figure 5.7 is the tropical tropospheric temperature (TTT) average from three satellite data products (NOAA, UAH and RSS) compared to the same layer average from climate models for 1979-2024. Again, the observed trends lie below the entire model range.

The wide range of choices made by modelers to characterize the physical processes in the models (see Box: Climate Modeling in Section 5.1 above) is seen by the large spread of trends in the middle troposphere, ± 40 percent about the median (Fig. 5.7). This vividly illustrates the uncertainties in attempts to model (parameterize) a complex system involving turbulence, moist thermodynamics, and energy fluxes over the full range of the tropical atmosphere’s time and space scales.

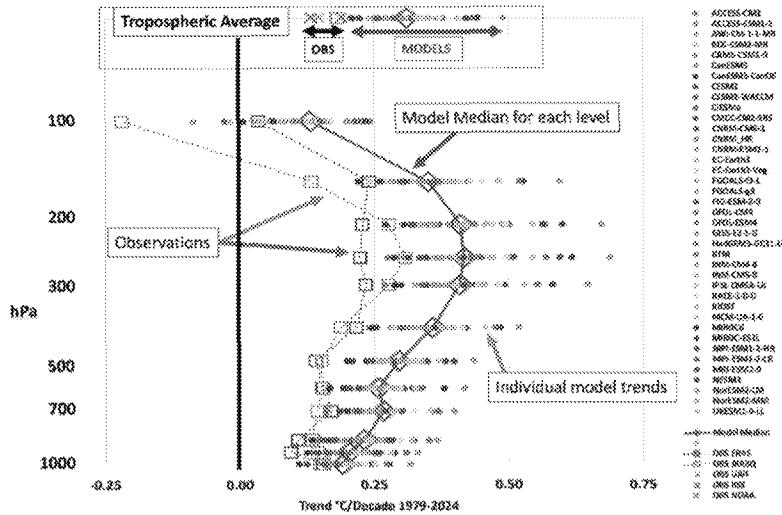


Figure 5.7: Modeled versus observed warming, tropical troposphere. Source: updated from Christy and McNider (2017) including data through 2024. Red line: model average. Green and blue lines: observational series (reanalysis).

This discrepancy has been the source of much controversy, with some arguing that even if there is very little observed warming aloft in the tropics, a “hotspot” still exists in the sense that the warming aloft is greater than at the surface (Santer et al. 2008). But there is good evidence that models also exaggerate the amplification rate. Klotzbach *et al.* (2009), showed that models project greater amplification with altitude than is observed. This result was subsequently confirmed by detailed time series analysis (Vogelsang and Nawaz 2016) which found that the model-observational difference is statistically significant.

The atmosphere’s temperature profile is a case where models are not merely uncertain but also show a common warming bias relative to observations. This suggests that they misrepresent certain fundamental feedback processes.

The IPCC AR6 did not assess this issue.

5.5 Stratospheric cooling

An important element of the expected general “fingerprint” of anthropogenic climate change is simultaneous warming of the troposphere and cooling of the stratosphere. AR6 acknowledged that cooling had been observed but only until the year 2000. The stratosphere has shown some warming since, contrary to model projections.

AR6 WG1 Ch 2 pp. 327-9 states:

Temperatures averaged through the full lower stratosphere (roughly 10–25 km) have decreased over 1980–2019 in all data products, with the bulk of the decrease prior to 2000. The decrease holds even if the influence of the El Chichon (1982) and Pinatubo (1991) volcanic eruptions on the trend,

found by Steiner *et al.* (2020a) to have increased the 1979–2018 cooling trend by 0.06°C per decade, is removed. Most datasets show no significant or only marginally significant trends over 2000–2019, and the results of Philipona *et al.* (2018) show weak increases over 2000–2015 in the very lowermost stratosphere sampled by radiosondes...

It is virtually certain that the lower stratosphere has cooled since the mid-20th century. However, most datasets show that lower stratospheric temperatures have stabilized since the mid-1990s with no significant change over the last 20 years. It is likely that middle and upper stratospheric temperatures have decreased since 1980, but there is low confidence in the magnitude.

The cited source, Philipona *et al.* (2018), in an article entitled “Radiosondes Show That After Decades of Cooling, the Lower Stratosphere Is Now Warming”, stated:

In response to continued greenhouse gas increases and stratospheric ozone depletion, climate models project continued tropospheric warming and stratospheric cooling over the coming decades. Global average satellite observations of lower stratospheric temperatures exhibit no significant trends since the turn of the century. In contrast, an analysis of vertically resolved radiosonde measurements from 60 stations shows an increase of lower stratospheric temperature since the turn of the century at altitudes between 15 and 30 km and over most continents. Trend estimates are somewhat sensitive to homogeneity assessment choices, but all investigated radiosonde data sets suggest a change from late twentieth century cooling to early 21st century warming in the lower stratosphere.

A combination of tropospheric warming and stratospheric cooling is a commonly cited “fingerprint” of anthropogenic warming. Stratospheric warming since 2000 coincides with continued surface and tropospheric warming, a pattern that is not found in climate model simulations and is not apparently consistent with the anthropogenic fingerprint.

5.6 Snow cover mismatch

Data compiled by the Rutgers University SnowLab show that Northern Hemisphere winter snow cover is not decreasing (Figure 5.8); if anything, it shows an increasing trend.

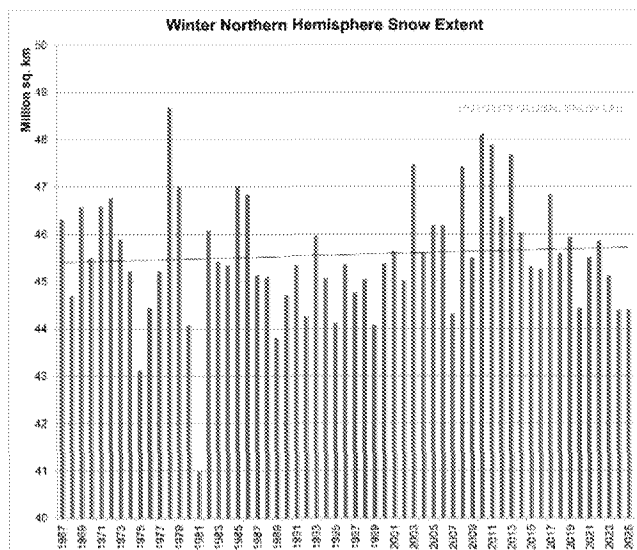


Figure 5.8: Northern Hemisphere Winter Snow extent.

Source: [[HYPERLINK](https://climate.rutgers.edu/snowcover/chart_seasonal.php?ui_set=nhland&ui_season=1)

"https://climate.rutgers.edu/snowcover/chart_seasonal.php?ui_set=nhland&ui_season=1"]
(accessed May 27, 2025)

Yet models project declining Northern Hemisphere snow cover in a warming climate, as described by Connolly *et al.* (2019).

The climate models were found to poorly explain the observed trends [in Northern Hemisphere snow cover]. While the models suggest snow cover should have steadily decreased for all four seasons, only spring and summer exhibited a long-term decrease, and the pattern of the observed decreases for these seasons was quite different from the modelled predictions. Moreover, the observed trends for autumn and winter suggest a long-term increase, although these trends were not statistically significant.

AR6 largely confines its discussion of changing Northern Hemisphere snow cover extent (SCE) to the Spring season, for which models and observations agree on a downward trend. Regarding Winter changes it remarks as follows (AR6 WGI Ch. 2 p. 344):

Assessment of SCE trends in the NH since 1978 indicates that for the October to February period there is substantial uncertainty in trends with the sign dependent on the observational product. Analysis using the NOAA Climate Data Record shows an increase in October to February SCE (Hernández-Henríquez *et al.*, 2015; Kunkel *et al.*, 2016) while analyses based on satellite borne optical sensors (Hori *et al.*, 2017) or multi-observation products (Mudryk *et al.*, 2020) show a negative trend for all seasons.

AR6 WGI Chapter 9 (p. 1284) points out that the NOAA Climate Data Record showing increased fall and winter SCE is inconsistent with land-based observations and model-based data sets. It notes that using optical satellite imagery to infer SCE in winter is challenging due to cloud cover and decreased solar illumination in winter months. Focusing on the Pacific Coast states (CA, OR and WA) cold season mountain snowfall that melts in spring and summer provides a substantial portion of warm season water resources. A comprehensive reconstruction of snowfall for the main source regions (Cascades and Sierra Nevada Mountains) indicates no significant trends in annual totals since the late 19th century (Christy 2022).

In summary, the up-to-date Rutgers SCE database indicates a mismatch between models and observations. Further work to reconcile conflicting trends in observational data sets is required.

5.7 Hemispheric symmetry of the planetary albedo

The planetary albedo is the fraction of incoming solar radiation reflected by the Earth back to space. It is an important element of the radiative energy balance and influences whether the planet will warm or cool over time. The planetary albedo is typically estimated at around 0.30; variations on the scale of 0.01 correspond to changes in solar forcing of about 3 W/m^2 , an amount larger than current anthropogenic forcing. It has long been noted that models disagree with each other and with observations on the value of the global planetary albedo (Stephens *et al.* 2015).

An intriguing property of the Earth's albedo is that, on average, the Northern Hemisphere (NH) and Southern Hemisphere (SH) have had nearly the same albedo, at least throughout the fifty-year satellite record (Stephens *et al.*, 2015). This symmetry is surprising, because the SH has much more ocean than land. Since ocean is less reflective than land, the NH should have higher albedo. Clouds (which are highly reflective) are more common in the NH and so compensate the surface albedo imbalances of the two hemispheres. Datseris and Stephens (2021) show that this cloud compensation comes from the extra-tropical storm tracks of the SH, which are cloudier than those of the NH. While the mechanism for this hemispheric symmetry is unclear, it likely operates on large temporal and spatial scales.

The hemispheric symmetry of the albedo is a simple gross metric for climate models. Rugenstein and Hakuba (2023) defined that metric as the difference between the NH and SH annual mean albedos, expressed as Wm^{-2} of the reflected sunlight, and compiled it for the CMIP6 climate models, as shown in Figure 5.9. Most of the CMIP6 models do not reproduce the small observed asymmetry (about 0.1 Wm^{-2}) and even disagree as to which hemisphere is more reflective. Moreover, the magnitude of the asymmetry ranges up to 5 Wm^{-2} in some models, twice as large as the current anthropogenic forcing (about 2.5 Wm^{-2}).

The significance of unphysical albedo asymmetries in the climate models is not yet fully known. However, other model studies suggest that interhemispheric changes in albedo can alter poleward heat fluxes, meridional temperature gradients, storminess, and differences in hemispheric ocean heat storage. The discrepancy between models and observations further raises issues regarding cloud feedback processes and so more generally diminishes confidence in model projections of the future climate.

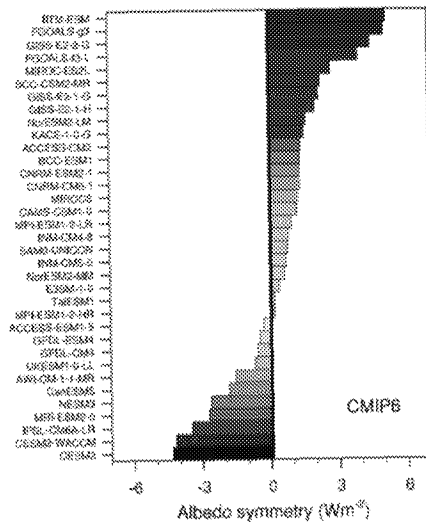


Figure 5.9. Differences in 20-year average reflectivity (albedo) between the Northern and Southern Hemispheres for CMIP6 models used in the most recent IPCC assessment (colored bars). The very small observed difference is indicated by the vertical black line. From Rugenstein and Hakuba (2023).

5.8 U.S. Corn Belt

One of the largest discrepancies between models and observations is in the U.S. Corn Belt, a region of particular importance to global food production. Figure 5.10 shows the warming trends for summertime (June, July, August) for the 12-state Corn Belt (IN, IA, IL, ND, SD, MO, MN, WI, MI, OH, KS, NE) during 1973-2022. All 36 climate models (red) warm far too rapidly compared to observations (blue).

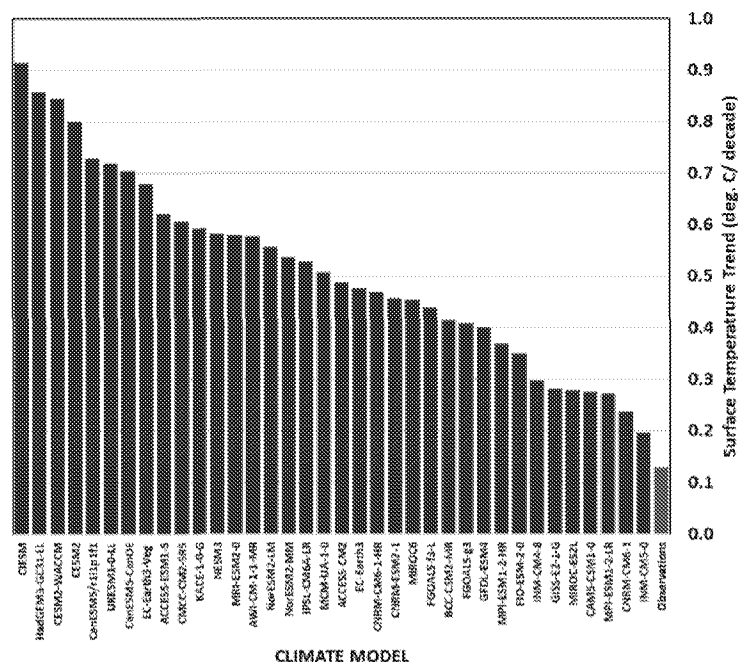


Figure 5.10: Modeled versus observed warming trends in the U.S. Corn Belt, 1973-2022.

As discussed in Chapter 9, the anticipated negative effects of increasing temperatures on U.S. corn yields have not materialized, in contrast to widely publicized studies proclaiming that theoretical future impacts are already being experienced (*e.g.*, Seager *et al.*, 2018).

The IPCC acknowledges limitations in the accuracy of regional climate model outputs. This example shows that users need to assess model projections carefully on a case-by-case basis since local biases might be sufficiently large that the models are simply not fit for purpose. As has recently been noted by two leaders of the modeling community (emphasis added)

... for many key applications that require regional climate model output or for assessing large-scale changes from small scale processes, we believe that **the current generation of models is not fit for purpose.** (Palmer and Stevens 2019)

To summarize:

- Climate models show warming biases in many aspects of their reproduction of the past few decades
- They produce too much warming at the surface (except in the models with lowest ECS), too much warming in the lower-and mid-troposphere and too much amplification of warming aloft
- They also produce too much stratospheric cooling, too much snow loss, and too much warming in the U.S. Corn Belt.

- The hemispheric albedo difference in individual climate models ranges widely in sign and magnitude compared to observations. The range in W/m^2 is three times larger than the direct anthropogenic forcing of CO_2 .
- The IPCC has acknowledged some of these issues but not others.

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6 EXTREME WEATHER

Chapter Summary

Most types of extreme weather exhibit no statistically significant long-term trends over the available historical record. While there has been an increase in hot days in the U.S. since the 1950s, a point emphasized by AR6, numbers are still low relative to the 1920s and 1930s. Extreme convective storms, hurricanes, tornadoes, floods and droughts exhibit considerable natural variability, but long-term increases are not detected. Some increases in extreme precipitation events can be detected in some regions over short intervals but the trends do not persist over long periods and at the regional scale. Wildfires are not more common in the U.S. than they were in the 1980s. Burned area increased from the 1960s to the early 2000's, however it is low compared to the estimated natural baseline level. U.S. wildfire activity is strongly affected by forest management practices.

6.1 Introduction

High impact weather extremes, usually related to temperature, precipitation and/or high wind, can disrupt infrastructure and therefore endanger human health and wellbeing. The issue is not whether extremes occur. Rather, it is whether there are long-term (decadal scale) changes in the frequency or character of extremes (“detection”), as well as the extent to which such changes and the attendant changes in hazards are caused by anthropogenic emissions of greenhouse gases (“attribution”; e.g., AR6 Seneviratne *et al.* 2021).

It is naïve to assume that any recent extreme event is caused by human influences on the climate. Climate is about the statistical properties of weather over decades, not single events. Further, there are only about 130 years of reliable observational records that can be analyzed statistically. That brief interval does not begin to contain all the extreme events that the climate system can create on its own. Over geologic time the climate system has generated an (essentially) infinite variety of weather patterns and extremes that humans have never observed and thus are absent from the databases used to determine extreme statistics [see *Perils of short data records* below]. For that reason, attributing an extreme event unprecedented in the record requires assumptions about the magnitude of natural variations.

This chapter is concerned with detection of trends in extreme weather, while Chapter 8 considers causal attribution, with Section 8.4 specifically addressing extreme weather. If no trend is detected, then clearly there is no basis for attribution. But even where a trend is observed, attribution to human-caused warming does not necessarily follow.

This is especially true for precipitation events. The hydrological literature has long noted the phenomenon of Long-Term Persistence (LTP) in rainfall data (Hurst 1951, Cohn and Lins 2005, Markonis and Koutsoyiannis 2016). LTP has a formal mathematical definition, but it can be thought of as the presence of natural variations with strong autocorrelations that result in cycle-like behavior on long time scales. The presence of LTP requires long records to accurately estimate variability. Analysis of records that are short relative to the LTP scale will tend to underestimate variability, therefore overstating the significance of apparent trends and underestimating the likelihood of extreme events (Cohn and Lins 2005).

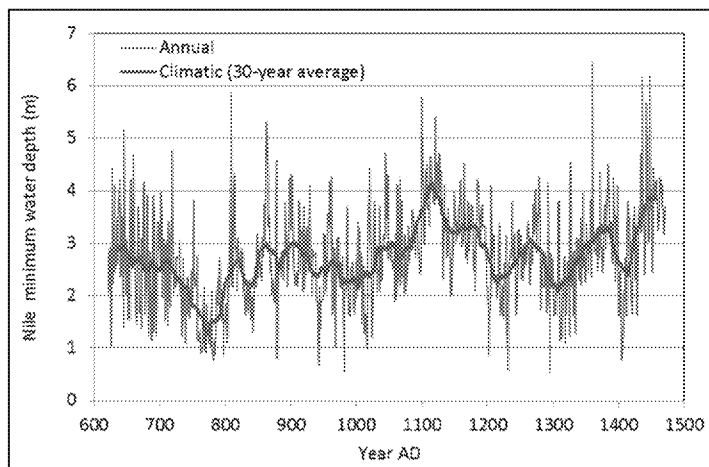


Figure 6.1.1: The annual minimum depth of the Nile River near Cairo over more than 650 years from 622 to 1284 A.D. The data, measured in meters, shows a characteristic pattern of year-to-year fluctuations around longer-term trends. Data from Koutsoyiannis (2013)

A good example of LTP is the eight-century long record of the annual minimum height of the Nile River observed at Roda Island in Cairo shown in Figure 6.1.1. The Nile River is fed by precipitation over a 4 million square mile drainage basin, an area equal to about one third of CONUS. Since human influences on the global climate were negligible before the 20th century, the century-scale variability of the thirty-year average is entirely natural; Egyptians of the seventh and eight centuries would have been incorrect to assume that the worsening drought during that time was the “new normal.”

With these caveats in mind, we examine the evidence for changes in selected weather and climate extremes. A recurring theme is the wide gap between public perceptions and scientific evidence. It has become routine in media coverage, government and private sector discussions, and even in some academic literature to make generalized assertions that extreme weather of all types is getting worse due to GHGs and “climate change.” Yet expert assessments typically have not drawn such sweeping conclusions and instead have emphasized the difficulty both of identifying specific trends and establishing a causal connection with anthropogenic forcing.

In the sections to follow we provide excerpts from various IPCC and NCA assessment reports denoting the sources as follows:

SREX: The IPCC Special Report on Managing the Risks of Extreme Events and Disasters to Advance Climate Change Adaptation (2012)

AR6: The IPCC Sixth Assessment Report Working Group 1 (2021).

NCA4: The U.S. Climate Science Special Report of the Fourth National Climate Assessment (2017) Volume I.

NCA5: The U.S. Climate Science Special Report of the Fifth National Climate Assessment (2023) Volume I.

In the excerpts, *italics* are in the original whereas **boldface** emphasis has been added.

Additionally we present updated data series from standard government sources to provide up-to-date information through 2024 wherever possible.

6.2 Hurricanes and tropical cyclones

AR6 provides the following assessment of tropical cyclones (TCs; used here as a synonym for hurricanes):

AR6: There is *low confidence* in most reported long-term (multidecadal to centennial) trends in TC frequency or intensity-based metrics due to changes in the technology used to collect the best-track data. (IPCC, 2021 p. 1585)

AR6: It is *likely* that the global proportion of major (Category 3–5) tropical cyclone occurrence has increased over the last four decades . . . There is *low confidence* in long-term (multi-decadal to centennial) trends in the frequency of all-category tropical cyclones. (IPCC, 2023 SPM p. 9)

AR6: A subset of the best-track data corresponding to hurricanes that have directly impacted the United States since 1900 is considered to be reliable, and shows **no trend in the frequency of U.S. landfall events**. (IPCC 2021 p. 1585)

Since 1980, when satellite observations first fully covered the global oceans, we have confidence in the numbers of total global hurricanes and major hurricanes (Category 3 and higher). Figure 6.2.1 shows that on average, each year there are about 50 hurricanes, with about 25 reaching major hurricane status (Maue, 2025). There is substantial year-to-year and decadal variability, a weak decrease in the number of hurricanes, and a slight but insignificant increase in the number of major hurricanes. These two trends combine to create an overall increase in the proportion of major hurricanes.

Global hurricane statistics are dominated by the Northwest Pacific Ocean, which accounts for ~35 percent of total global hurricanes, whereas the Atlantic accounts for ~15 percent of global hurricanes (Colorado State University, 2025). Data in the Atlantic basin extends further back than in the other ocean basins and is most relevant to U.S. policy makers.

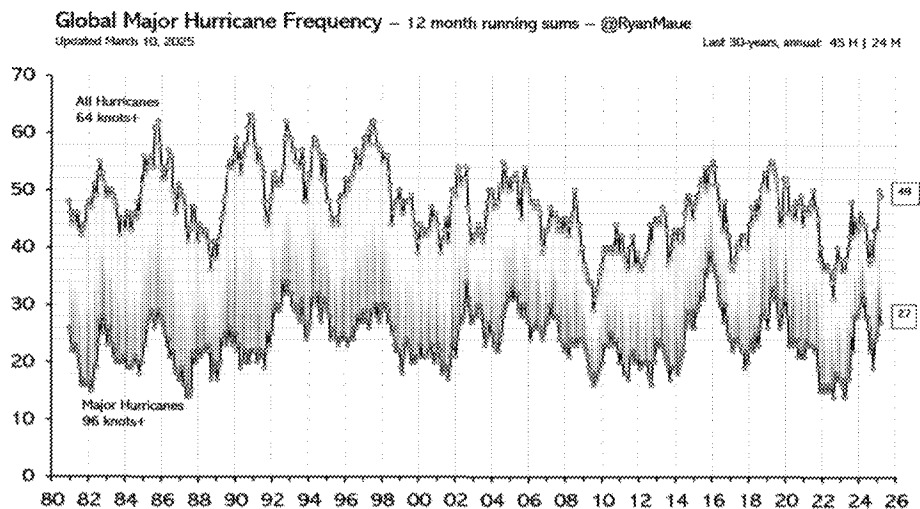


Figure 6.2.1: Global frequency of hurricanes and major hurricanes since 1980.

Figure 6.2.2 shows the frequency of Atlantic hurricanes and major hurricanes (Category 3 and higher) back to 1920. Data prior to 1965 (the onset of satellite observations in the Atlantic, shaded in blue) shows some undercounting, with data prior to 1920 showing substantial undercounting (Vecchi and Knutson, 2011). All measures of Atlantic hurricane activity show a significant increase since 1970. However, the period from 1971-1994 saw exceptionally low activity, with high activity (comparable to the past two decades, even with undercounting) also observed during the 1950's and 1960's, and even in the 1930's.

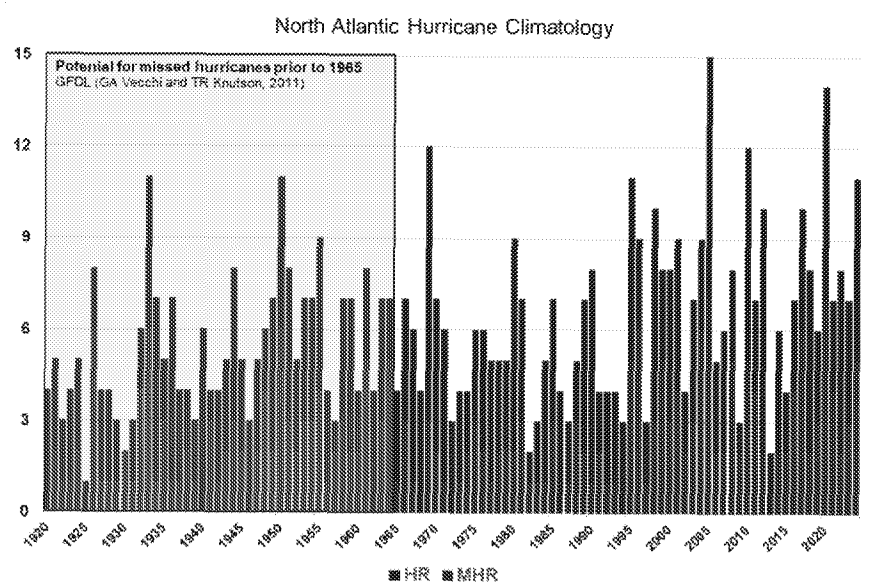


Figure 6.2.2: Atlantic frequency of hurricanes (HR) and major hurricanes (MHR) since 1920. Source: National Hurricane Center (2024)

Figure 6.2.2 shows that Atlantic hurricanes vary strongly on decadal and multidecadal time scales. These variations are associated primarily with the Atlantic Multidecadal Oscillation (AMO), which is manifest in basin-wide sea surface temperature and sea level pressure fluctuations connected to large scale ocean circulation patterns. The AMO was in its warm phase during 1926-1970 and 1995-present, but in its cool phase during 1971-1994. It has its greatest impact on the number of major hurricanes (Category 3+), identified by Goldenberg *et al.* (2001) as associated with above normal SSTs and decreased vertical shear in the AMO warm phase (see also Bell and Chelliah, 2006; Klotzbach *et al.*, 2018).

Klotzbach *et al.* (2018) conducted a comprehensive evaluation of the landfalling hurricane data for the Continental U.S. since 1900. Figure 6.2.3 updates their analysis through 2024. While the largest numbers of landfalling hurricanes are from 2004, 2005 and 2020, there is no statistically significant trend since 1920. Figure 6.2.3 also shows the time series for major hurricane landfalls (Category 3-5). The largest year in the record is 2005, with 4 major hurricane landfalls. However, following 2005 there were no major hurricanes striking the U.S. through 2016, the longest such period since 1920.

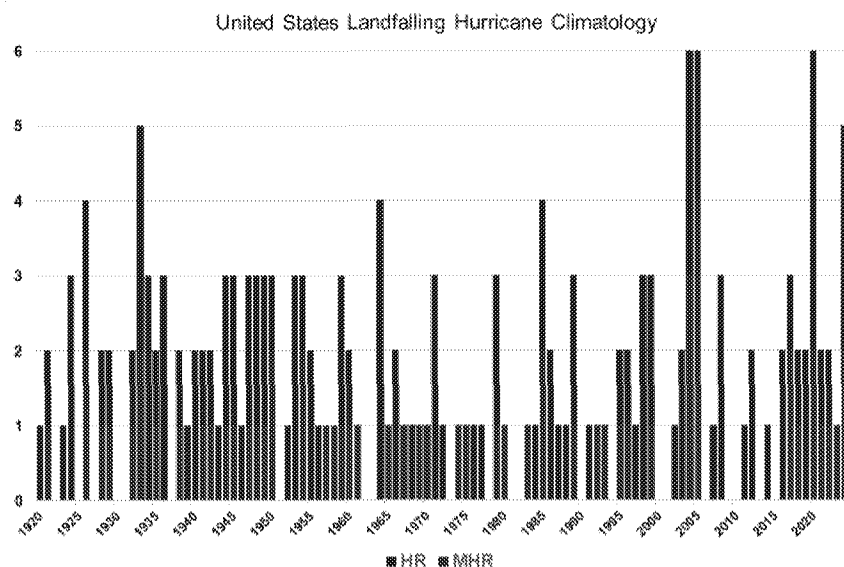


Figure 6.2.3: U.S. landfalling frequency of hurricanes (HR) and major hurricanes (MHR) since 1920. Source NOAA HRD(a) (2024)

Figure 6.2.3 shows substantial interannual to multidecadal variability in U.S. landfall activity. Klotzbach *et al.* (2018) examined how the landfall counts vary with ENSO (El Niño versus La Niña) and the warm versus cold phases of the Atlantic Multidecadal Oscillation (AMO).

Villarini *et al.* (2012) provide an analysis of U.S. hurricane landfalls back to 1878. While it is possible that some landfalls were missed in the late 19th century owing to sparsely populated regions on the Gulf Coast, it is remarkable that the highest year in the entire record is 1886, with 7 hurricane landfalls, when human influences on the climate were much smaller than they are today.

Table 6.2.1 shows the 10 strongest hurricanes (plus ties) to make U.S. landfall. Of the hurricanes that have made landfall with sustained winds greater than 150 mph, only one has occurred in the 21st century.

In summary, analyses of both global and regional variability and trends of hurricane activity provide the basis for detecting changes and understanding their causes. The relatively short historical record of hurricane activity, and the even shorter record from the satellite era, is not sufficient to assess whether recent hurricane activity is unusual relative to the background natural variability. Atlantic hurricane processes are influenced substantially by the natural modes of ocean circulation variability in the Atlantic, notably the Atlantic Multidecadal Oscillation. While it has long been hypothesized that a rising global sea surface temperature would cause an increase in hurricane intensity, identification of any significant trend in the hurricane data is hampered by a short data record and substantial natural variability.

Rank	Year	Landfall Wind (MPH)	Name
1	1935	185	"Labor Day"
2	1969	175	Camille
3	1992	165	Andrew
4	2018	160	Michael
5	1856	150	"Last Island"
5	1886	150	"Indianola"
5	1919	150	-----
5	1932	150	"Freeport"
5	2004	150	Charley
5	2020	150	Laura
5	2021	150	Ida
5	2022	150	Ian

Table 6.2.1 Strongest hurricanes to make landfall along the U.S. coast. Source (NOAA HRD(b), 2024)

6.3 Temperature extremes

The AR6 assessment focused on the period after 1950 and reported increasing trends in heatwave frequency and intensity. However, NCA4 noted that heatwave activity in the U.S. reached a peak in the 1930s (Figure 6.3.1).

AR6: It is *virtually certain* that hot extremes (including heatwaves) have become more frequent and more intense across most land regions since the 1950s, while cold extremes (including cold waves) have become less frequent and less severe (SPM, A3.1)

AR6: In North America, there is very robust evidence for a *very likely* increase in the intensity and frequency of hot extremes and decrease in the intensity and frequency of cold extremes for the whole continent, though there are substantial spatial and seasonal variations in the trends. Minimum temperatures display warming consistently across the continent, while there are more contrasting trends in the annual maximum daily temperatures in parts of the USA. (Chapter 11, p 1550)

NCA4: Changes in warm extremes are more nuanced than changes in cold extremes. For instance, the warmest daily temperature of the year **increased** in some parts of the West over the past century, but there were **decreases** in almost all locations east of the Rocky Mountains. In fact, all eastern regions experienced a net **decrease**, most notably the Midwest (about 2.2°F [1.2°C]) and the Southeast (roughly 1.5°F [0.8°C]). (pp. 190-191)

NCA4: Since the mid-1960s, there has been only a very slight increase in the warmest daily temperature of the year (amidst large interannual variability). Heat waves (6-day periods with a maximum temperature above the 90th percentile for 1961–1990) increased in frequency until the mid-1930s, became considerably less common through the mid-1960s, and increased in frequency again thereafter. As with warm daily temperatures, **heat wave magnitude reached a maximum in the 1930s.** (pp. 190-191)

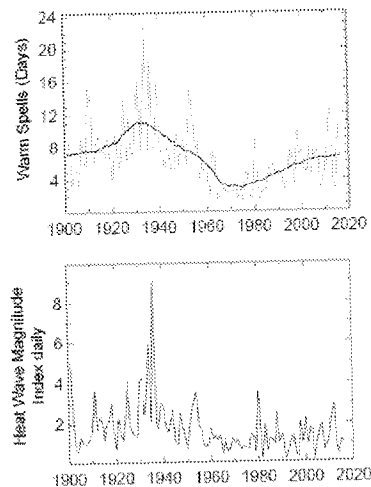


Figure 6.3.1: U.S. Heat waves since 1900. Source: NCA4 Figure 6.4

6.3.1 Temperatures in the U.S. are becoming less extreme

Daily maximum temperatures in the warm season (T max, May-Sep) and daily minimum temperatures in the cold season (T min, Dec-Mar) are available beginning in Dec 1898 (126 years). The dataset consists of 1,211 CONUS stations designated as United States Historical Climate Network or USHCN stations (see Figure 6.3.2; Quinlan *et al.* 1987, Karl *et al.* 1990). These stations were selected by NOAA as having the fewest problematic issues with gaps, station moves, and instrument changes. Where gaps still exist, nearby stations (bias-corrected) were merged so that the median volume of data available for a station is 98%. Although there are certainly errors in the dataset, including unresolved spurious warming due to UHI effects that especially bias T min records, this data set is sufficiently accurate for assessing trends in T max heat extremes.

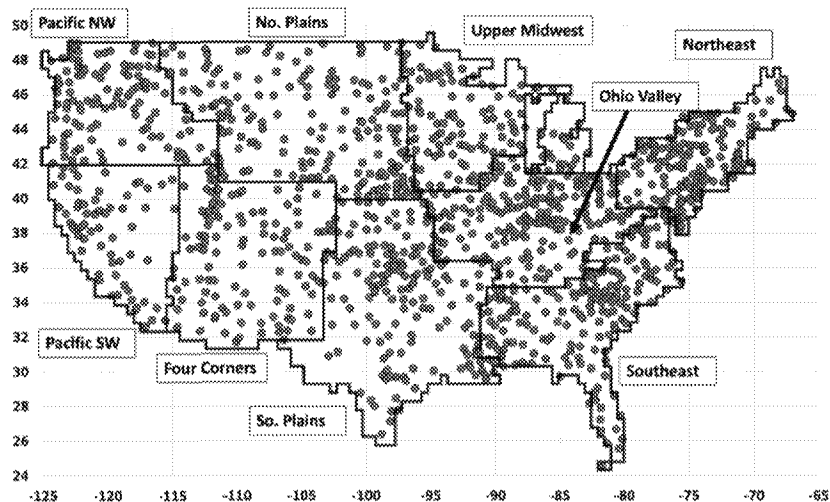


Figure 6.3.2 Locations of USHCN temperature stations. Source: USHCN.

We begin with the question of whether the occurrence of daily record high or low temperatures has changed since Dec 1898. Each warm-season has 153 days (1 May to 30 Sep) and each cold-season has 122 days (1 Dec to 31 Mar). For each station and day, we calculated the year in which the record highest (lowest) temperature occurred. With 126 years of observations, if there were no temperature trends over time, the expected number of records for Tmax would be 1.21 ($=153/126$) per station per year and for Tmin 0.96 ($=122/126$) per station per year.

Figure 6.3.3 shows the observed distribution in time of the occurrence of these extreme events. There is a common feature in many metrics of warm-season extremes in the CONUS - the exceptional heat of the 1920s and especially the 1930s, peaking in 1936. On a per-station average, 60 percent of the Tmax records and 59 percent of the Tmin records occurred in the first half of the period (1899-1961).

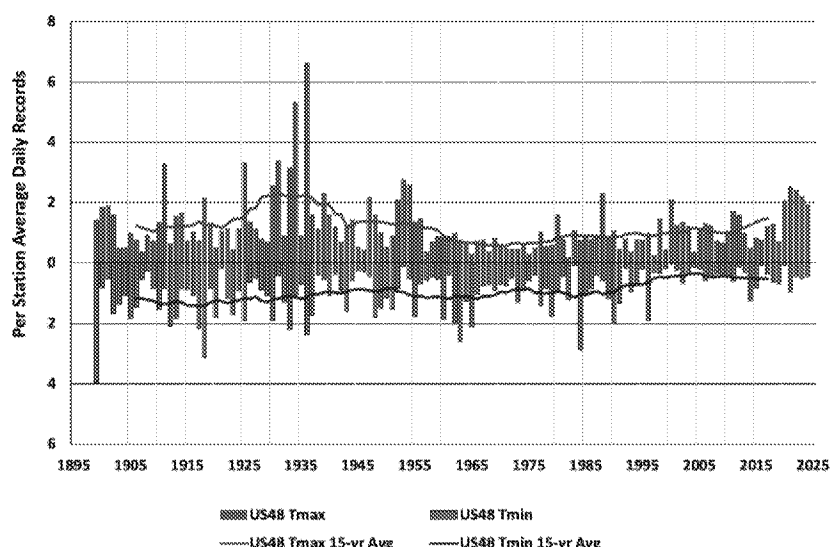


Figure 6.3.3 Number of daily record High (red) and Low (blue) temperatures for warm and cold seasons in the CONUS. The lines represent the 15-year running, centered average. Source: 1,211 USHCN stations supplemented as needed to achieve a minimum of 92 percent of observations in the 126-year period since Dec 1898. US48: contiguous U.S. states. Tmax: maximum temperature. Tmin: minimum temperature.

On the cold side, the Valentine's Day Arctic outbreak in Feb 1899 stands as the most extensive cold extreme experienced by CONUS, with 1917 in 2nd place. The frequency of cold records has declined, especially over the last quarter of the period in which only 13 percent of the extreme cold events were measured. In contrast, 25 percent of the extreme Tmax records were achieved in the last quarter, in accordance with statistical expectations. These general features have been noted in past assessments (see above, IPCC AR6, NCA4). Combining the two histories, the overall reduction in numbers of both cold and hot extremes over the past century indicates a climate less prone to extremes.

This pattern is also shown in Figure 6.3.4. For each station and for each year the hottest warm season and coldest cold-season temperatures were calculated. Then the differences between these were computed by station and geographically-averaged over all stations, thus yielding an annual measure of the expected range of local extreme temperatures for each year. Figure 6.3.4 shows the 15-year trailing average of this measure, which has clearly declined over the past century.

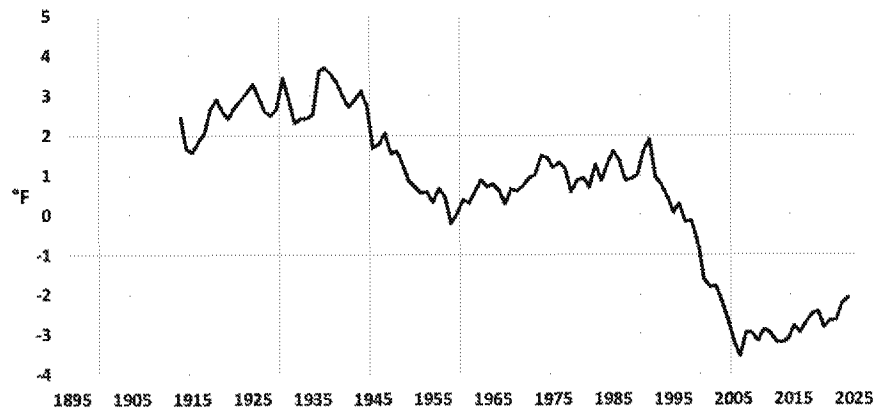


Figure 6.3.4. 15-year trailing average of the difference each year of each station's hottest warm-season Tmax and coldest cold-season Tmin relative to the long-term average. Source: Author analysis of USHCN data.

The average difference for each station between the hottest summer Tmax and coldest winter Tmin has declined by about 5°F in the past 126 years. The decline is due mostly to warmer winter Tmin, but a decline in summer Tmax is also a factor. The rise in Tmin has been strongly related to the growing presence of manufactured surfaces around the weather stations over the last 100+ years (the so-called urban heat island effect; Section 3.3 and Spencer *et al.* 2025).

In summary, while temperature extremes are regularly experienced in the U.S. and attract a great deal of media attention, long term records show the U.S. climate has become less extreme over time (milder) when measured by the range between warm season maxima and cold season minima.

6.3.2 Exceedances of a heat threshold

Under the heading of “The Risk of Temperature Extremes is Changing”, the most recent U.S. National Climate Assessment report (NCA5) notes the increase in a threshold metric of number of days at or above 95°F, stating,

The western U.S. has been particularly affected by extreme heat since the 1980s ..., experiencing a larger increase in days over 95°F, as would be expected given the greater warming in that region relative to the eastern US. Several major heatwaves have affected the U.S. since 2018, including a record-shattering event in the Pacific Northwest in 2021.

Are the occurrences of 95°F days changing? In a climate as varied as that of CONUS, threshold statistics can be misleading. A region with many stations that have near 95°F Tmax average temperatures in the summer might see large swings in the metric when only small changes in average temperature occur. Elsewhere with stations that either rarely or virtually always achieve 95°F Tmax temperatures, a small change will not have much impact on the results.

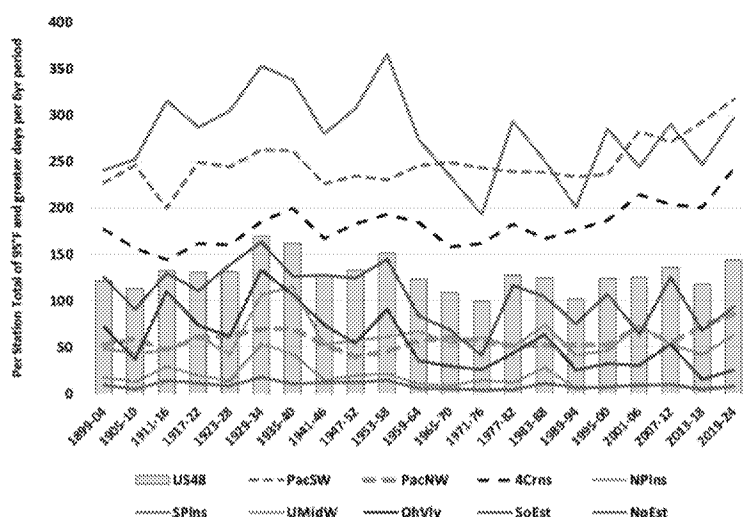


Figure 6.3.5. Total number of days $\geq 95^{\circ}\text{F}$ in 6-yr periods, U.S. 48 (bars) and regions (lines). A 6-year period is used as this evenly divides the 126-year record. US48: contiguous U.S. states. See Figure 6.3.2 for region names. Source: Author analysis of USHCN data.

In the past 126 years, the average CONUS station experienced 129 days exceeding 95°C per 6-yr period, but the regional values range from 278 in the Southern Plains to 9 in the Northeast. Thus, such threshold analyses must be interpreted with caution. Figure 6.3.5 shows that only three of the nine regions, all in the West, have experienced upward trends in the number of 95°F or hotter days (dashed lines). The CONUS as a whole has not, and the other six regions have experienced declines.

The Pacific NW heatwave of 2021 referenced in the NCA5 quote will be examined more closely in Section 8.6.1. The evidence indicates that it was a single, unprecedented event in the record, not part of a pattern of increasing extreme heat. For example, the 5-day average tropospheric grid-point temperature anomaly over the Pacific NW during that event was $+10.8^{\circ}\text{C}$, the most extreme Northern Hemisphere grid point summer anomaly in the 46 years from over 4 million grid values. In contrast, the global temperature anomaly during that time was virtually zero ($+0.03^{\circ}\text{C}$, Mass *et al.* 2024).

The NCA5 heading “The Risk of Temperature Extremes is Changing” suggests pervasive positive trends are now being observed in threshold temperatures, but that is not evident in the figures above.

6.3.3 Heatwaves

Heatwaves (consecutive days that exceed an extreme threshold) have a greater societal impact than a single daily record temperature. We measure “Heatwave Days” here as the count of all days in May-Sep each year that exceed the 90th percentile for that day and that lie within a period of at least six consecutive days. This is equivalent to the method used in NCA4, except that the reference period here is the entire record 1899-2024 while NCA4 truncated the reference period to 1961-1990, which was a cool period interval. That truncation boosts positive results (days exceeding the 90th percentile) in periods years warmer than the reference period, especially starting in 1960 and moving to the present (see Figure 6.3.6 and discussion below).

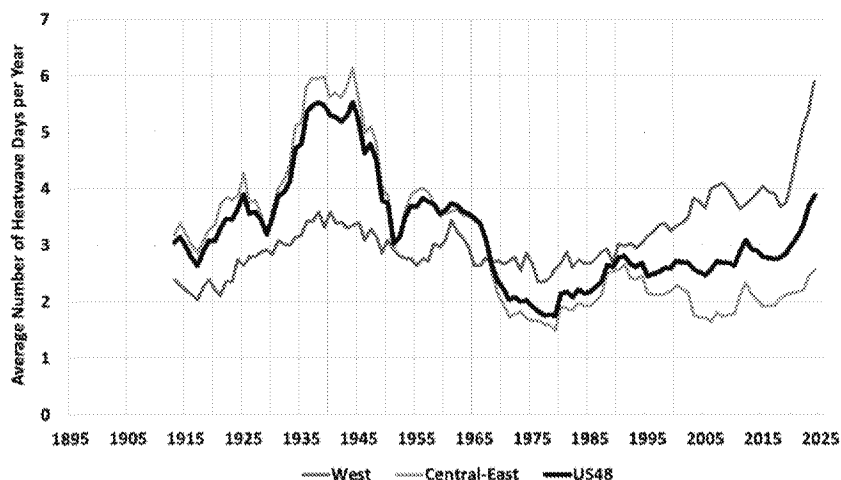


Figure 6.3.6 15-year trailing average of number of heatwave days per year per station in the CONUS (blackline) and two regions: West (red), Central-east (green).

Figure 6.3.6 indicates that there are regional variations in heatwave activity. The excessive heat of the first half of the 20th century occurred primarily in the eastern two-thirds of the nation, while the West has seen a recent increase of heatwave days (NCA5). This indicates that the background warm season circulation favored heatwaves in the eastern portions of the country in the first half of the 20th century, but in the 21st century the patterns have favored heatwaves in the West. For CONUS as a whole, heatwaves are no more common today than they were a century ago, consistent with the upper panel of our Figure 6.4.1 taken from NCA4.

This metric varies significantly with region. The four northern regions (Pacific NW, Northern Plains, Upper Midwest, and Northeast) on average experience 15 to 27 heatwave days per 15-year period. In contrast, the five southern regions (Pacific Southwest, 4-Corners, Southern Plains, Ohio Valley and Southeast) see 37 to 54 such days, essentially twice as many. This suggests the summer circulation pattern is more prone to stationary events in the southern regions while transient systems in the northern regions are more common and thus cut short these potentially longer events.

The analysis of heatwaves is an example of why it is important to consider complete datasets and appropriate metrics. The NCA5 directs readers to the website [[HYPERLINK "https://www.globalchange.gov/indicators/heat-waves"](https://www.globalchange.gov/indicators/heat-waves)] (USGCRP 2023) to view a figure showing the number of urban heatwaves by decade from the 1960s, which we reproduce as Figure 6.3.7.

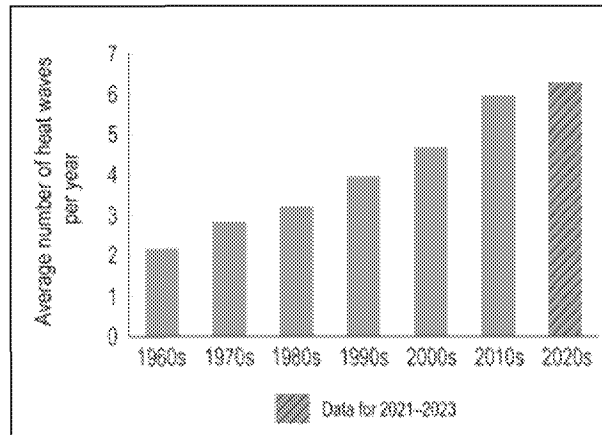


Figure 6.3.7: Average number of urban heatwaves per year for 50 large U.S. metropolitan areas a misleading metric for reasons explained in the text. From [HYPERLINK "https://www.globalchange.gov/indicators/heat-waves"] (accessed May 22, 2025).

The figure shows a monotonic increase in each decade from two occurrences per year in the 1960s to six in the 2020s. The definition of a heatwave used is an unusual but practical measure of human discomfort - a period of at least 2 consecutive days when the *minimum* apparent temperature (combination of temperature and humidity) exceeds the 85th percentile. Note too, the dataset is limited to the 50 largest U.S. cities.

Given the unusual heatwave definition and urban focus, these increasing values since 1960 presented in USGCRP (2023) are not informative about long term trends or the influence of GHG emissions for at least two reasons. First, as shown in Figures 6.3.1 and 6.3.6, the 1960s was the coldest decade and 1970s the second coldest decade since the 1910's, so this starting date preconditions the time series to show increases. Second, post-1960 urbanization in these cities is a major factor in the rise of T_{min} relative to co-located T_{max} and relative to T_{min} at nearby rural stations. This does not dismiss the real rise in nighttime temperatures in major U. S. cities and the societal impacts associated with these changes. However, we note for a variety of reasons that summer T_{max} (especially in rural areas) is a better metric for detecting changes in heatwaves influenced by changes in the background climate due, for example, to increasing GHGs (Christy *et al.* 2009). For CONUS as a whole, the evidence in this section suggests GHG emissions have had little-to-no effect on heatwaves against the background of urbanization and natural climate variability.

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6.4 Extreme precipitation

AR6 assessed that an increase in heavy precipitation has been observed in data starting in the 1950s.

AR6: The frequency and intensity of **heavy precipitation events have increased** since the 1950s over most land area for which observational data are sufficient for trend analysis (high confidence). (SPM A3.2)

AR6: In North America, there is robust evidence that the **magnitude and intensity of extreme precipitation has very likely increased** since the 1950s. Both [one-day maxima] and [5-day maxima] have significantly increased in North America during 1950-2018. (Chapter 11, p. 1560)

The U.S. National Climate Assessments (NCA4, NCA5) have highlighted an increase in the occurrence of the heaviest precipitation events (defined in different ways) primarily in the eastern half of CONUS, especially the Northeast, when starting the analysis in either 1901 or 1958. Interestingly, the regional variations indicate that the largest increases in extreme precipitation events are in the Northeast and the smallest in the West, a pattern counter to the changes in temperature extremes (Figure 6.3.6).

McKittrick and Christy (2019) examined long-term and consistent station observations of extreme daily precipitation to test some of these NCA claims for the Southeast and West Coast using a trend model with a non-parametric variance estimator robust to the complex autocorrelation properties of precipitation data. When the time series were extended back in time (as far as 1872 in some cases) or started later (1978), there were no significant trends for either region.

These findings have been updated for this report (McKittrick and Christy 2025) with similarly constructed observations from 29 stations on the CONUS Pacific Coast (1893ff from San Diego CA to Blaine WA) and 24 stations in the humid Southeast (1872ff from Austin TX to Washington DC), also adding 27 stations in the Northeast (1888ff from Buffalo NY to Eastport ME). The locations are shown in Figure 6.4.1. The stations were selected based on availability of long-term high-quality records. The regions are each associated with important features of extreme precipitation behavior: the Pacific coast is associated with landfalling atmospheric rivers for which AR6 cites evidence of increasing activity since 1948 with further increases expected as the world warms (AR6 8.3.2.8.2); the NCA report indicates that the Northeast has experienced the greatest increase in extreme events, and the Southeast is also a place noted in the NCAs as having increased extreme events.

The results of applying the analysis of McKittrick and Christy (2019) were as follows for each region, then followed by further explanation.

Pacific Coast heavy rainfall events

- The average precipitation trend is statistically significant (downwards) in Astoria OR; insignificant elsewhere.
- The trend in rainfall variance is positive and significant in Big Sur CA; insignificant elsewhere.
- The trend in daily maximum precipitation is positive and significant in Aberdeen WA and Big Sur CA and negative and significant in Newport OR (insignificant elsewhere).
- Averaged over all stations in the region, none of these three trend parameters is statistically significant.

The Pacific Coast receives considerable precipitation from Atmospheric River (AR) events which often last more than a day or two (e.g., Gershunov *et al.* 2017, Pan *et al.* 2024). The worst series of such events in recent history was the so-called ARkStorm that occurred between December 1861 and January 1862; it dumped nearly 10 feet of rain in parts of California and submerged the entire Central Valley for weeks under as much as 15 feet of water (Brewer 1930, Null and Hulbert 2007). Additionally, Paleoclimate research has found six megastorms more severe than 1861–1862 in California during the last 1800 years, occurring at intervals of 300 years or so (Porter *et al.* 2011).



Figure 6.4.1: Locations of precipitation monitoring stations used in this report. Orange: Pacific coast. Blue: Northeast. Green: Southeast. Data from McKittrick and Christy (2025).

We examine occurrences of 5-day deluges as follows. Taking the Pacific coast as an example, a 130-year span contains 26 5-year intervals. At each location we computed the 5-day precipitation totals throughout the year and selected the 26 highest values across the sample. A single year might have more than one of the 26 heaviest. Each of those can be thought of as 1-in-5yr events. If there are no trends in precipitation, then the total number of these events across all stations should be evenly spread over the years. In Fig. 6.4.2 we show the distribution in time of these events for the Pacific coast. The deluges associated with the massive 1997-98 El Niño event are readily apparent. While erratic, as is typical of such precipitation metrics, there is no indication of a tendency to become more frequent over time.

Southeast heavy rainfall events

- The trend in average precipitation is positive and statistically significant in Mobile AL and Quitman GA but insignificant elsewhere.
- The trend in rainfall variance is positive and significant in Mobile AL but insignificant elsewhere.
- The trend in daily maximum precipitation is positive and significant in Vicksburg MS and Norfolk VA but insignificant elsewhere.
- Averaged over all stations in the region none of these three trend parameters is statistically significant

Figure 6.4.3 is analogous to Figure 6.4.2 for the last 150 years in the Southeast humid zone. The temporal pattern of 5-day totals of the 1-in-5yr heavy events is generally unremarkable, though a cluster of higher values appears in 1995 to 2019. The increase in those years is due largely to the 4 northeastern-most stations of Wilmington NC, Weldon NC, Washington DC, and Norfolk VA. This confirms the pattern indicated in NCA4 and NCA5 -- an increasing frequency of heavy events due to a temporal clustering of tropical storms from eastern NC to Maine discussed below. Otherwise, the remaining 20 stations show an unremarkable temporal distribution of heavy events.

BOX: Perils of short data records

San Francisco provides a good case study of the limitation of using short historical samples to characterize natural variability of extreme events. Suppose we use a 130-year sample of daily San Francisco precipitation from 1895 to 2024 and we look for 3-day, 5-day, 14-day and 30-day rainfall records. The results are as shown in Table 7.2.

Event	Record (inches)	Year
3-day	6.94	2023
5-day	8.55	2023
14-day	12.62	2023
30-day	18.93	1998

Table 6.2: Extreme rainfall records, San Francisco, 1895-2024.

The records all cluster in the more recent years. 2023 appears to be an exceptional year and since it is near the end of the sample, it might suggest that the climate has shifted into a more hazardous state, perhaps because of human influences. [Note “2023” indicates the event occurred in the water-year of Aug 2022 to Jul 2023.]

But the picture is very different if we use a sample that begins 45 years earlier, in 1850. Table 6.3 shows that the record-setting events all happened in the 1860s. Furthermore 2023 is now not even in 2nd place but falls to 3rd or 4th place. And comparing the records of the two charts shows the extreme precipitation events in 1862 and 1867 involved considerably more rainfall than the 1998 and 2023 events, with 14- and 30-day totals about 50 percent higher.

Event	Record (inches)	Year	Rank of 130-yr extreme listed in Table 6.2
3-day	8.85	1867	3
5-day	9.80	1867	3
14-day	19.05	1862	4
30-day	28.25	1862	2

Table 6.3: Extreme rainfall records, San Francisco, 1850-2024.

The range of natural variability is made even more remarkable when paleoclimate evidence is examined. Porter *et al.* (2011) discovered that in the past 1,800 years at least six megastorms were more intense than the devastating 1861-62 ARkStorm that struck the region. Evidently such extreme events, “unprecedented” in our 1895-2024 sample, have impacted the region about every 300 years, though not since 1895.

This example illustrates the limitations of using relatively short climate periods (~130 years) to assess the character and range of natural variability in general and of extreme events in particular. Accurate representation of the full range of natural variability is necessary for any attribution analyses (Section 8.6). Infrastructure planners, emergency management institutions and attribution scientists would understand the significant mischaracterization of the magnitude of a future extreme if based only on the last 130 years. In this case, a single time-sample of 130 years provides an underestimation of the extreme value by up to 50 percent determined when adding only 45 more years of observations. Compared to millennial-scale paleoclimate evidence, an even greater underestimation would occur. An important lesson is that the climate can deliver great surprises on its own, even without human influences.

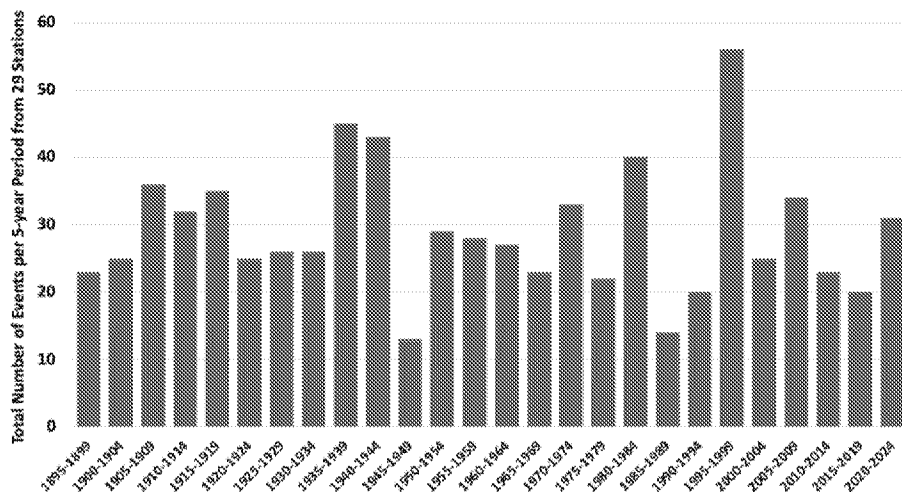


Figure 6.4.2. The time distribution by 5-year periods of the 26 heaviest (1-in-5 yr) occurrences for 29 stations on the Pacific coast.

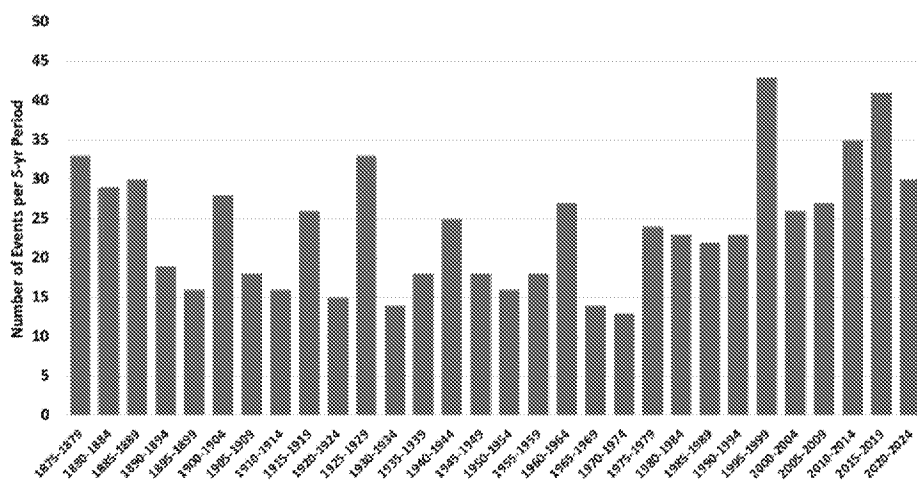


Figure 6.4.3. As in Fig. 6.4.2 but for the heaviest 30 (1-in-5yr) events for 24 stations in the humid Southeast from Austin TX to Washington DC in 5-year bins for 1875-2024.

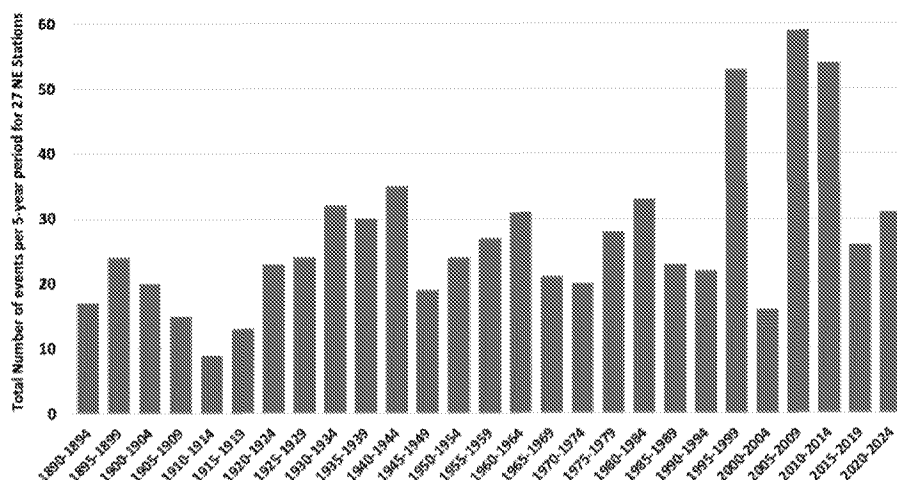


Figure 6.4.4. As in Fig. 6.4.2 but for the heaviest 27 (1-in-5yr) 3-day precipitation events for 27 Northeast stations from NY to ME, including Montreal.

Northeast heavy rainfall events

- The trend in average precipitation is positive and statistically significant in 12 of 27 locations and also in the regional average.
- The trend in rainfall variance is positive and significant in Portland ME, Albany NY, Buffalo NY and Eastport ME but insignificant elsewhere.
- The trend in daily maximum precipitation is positive and significant in Portland ME, Gardiner ME and Eastport ME but insignificant elsewhere.
- When averaged over all stations in the region, there is no statistically significant trend in either the precipitation variance or maximum

Fig. 6.4.4 is analogous to Figure 6.4.2 for the last 135 years in 27 stations in the Northeast (including Montreal Canada). We use 3-day totals here as this produced the largest temporal variations in time. In this region, 77 percent of events occur during June to October and are dominated by incursions of hurricanes, tropical storms, or tropical storms that transition to extratropical systems. According to NCA4 and NCA5 this region experienced the largest increases in extreme events, so it merits a closer examination.

There is a noticeable clustering of extreme events from 1995 to 2014. Howarth *et al.* (2019) examined a similar region as in Fig. 6.4.4 that includes PA and NJ and reported significant differences in various precipitation extremes between two 18-year periods, 1979-96 and 1997-2014. That included a 317 percent increase in 24-hr events exceeding 6 inches, while we find a 58 percent increase over the same years. However, Figure 6.4.4 shows that frequency drops sharply after 2014, returning to the long-term average in the subsequent 5-year intervals, again illustrating the perils of drawing conclusions from short-term trends in highly variable metrics.

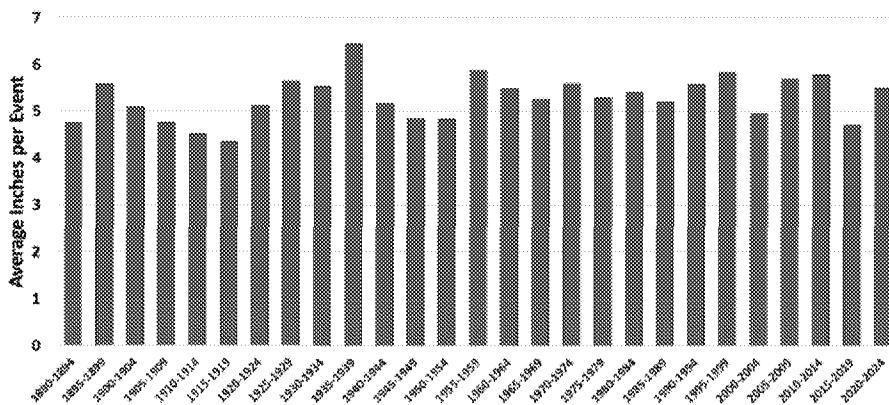


Figure 6.4.5. Average amount of precipitation falling in the 1-in-5yr events for the NE stations

The high percentage increase in the Howarth *et al.* sample is associated with small numbers at relatively few locations: there were only 6 in the first period and 25 in the second across 58 stations. Most of the stations did not record such an event. If there is a region-wide increase in heavy events, it should be seen in the average across all stations. Figure 6.4.5 shows the average precipitation in the 1-in-5 yr events across the NE. The trend is only +0.04 inches/decade. The highest amount, 1935-39, includes the Great New England Hurricane of 1938, one of the rare major (Category 3 or higher) hurricanes to strike the region. The results in Figs. 6.4.4 and 6.4.5 thus suggest that though there was a surge in the number of events at a few locations during the 1995 to 2014 interval, there was no regional pattern and the change did not persist beyond 2014.

Jong *et al.* (2024) document the increase of tropical influences on precipitation events in the Northeast since 1959 and concluded “The autumn extreme precipitation trend over the Northeast U.S. is primarily attributed to tropical cyclone-related events since the 1990s.” The question then becomes: “Was the temporal clustering of tropical systems in 1997-2014 which affected the Northeast a response to increasing GHGs?” Jong *et al.* examined CMIP-6 model output which suggests that there will be fewer such systems in the 21st century but that the intensity of the rainfall events might increase. This conjecture is not seen in Fig. 6.4.5 where the amount-per-event has remained steady over the 135-year period.

There is some evidence to indicate the heaviest rainfall events might be redistributed due to the impact of urban infrastructure on the local weather (e.g., Pielke Sr. *et al.* 2011, Zhang *et al.* 2018, Yang *et al.* 2024). Yang *et al.* state “Cities that experience compact development tend to witness larger increases in extreme rainfall frequency over downtown than their rural surroundings, while the anomalies in extreme rainfall frequency diminish for cities with dispersed development.” While this is an important insight to consider, the effect on the specific stations used in this analysis is unknown, or at least not detectable in Fig. 6.4.5.

In summary, some U.S. regions show short-duration increases in extreme precipitation events, consistent with natural variability. But analysis of long term, nationwide historical records that considers the autocorrelation properties of precipitation data does not support the claim that extreme short-duration rainfall events are becoming more frequent or intense.

6.5 Tornadoes

AR6 assesses tornado trends in the U.S. as follows:

[O]bservational trends in tornadoes, hail, and lightning associated with severe convective storms are not robustly detected due to insufficient coverage of the long-term observations. There is *medium confidence* that the mean annual number of tornadoes in the USA has remained relatively constant. (Chapter 11, section 11.7.3, p. 1594)

The monitoring of weak tornadoes has changed over time. The growth of rural populations and the increasing ability to take video with hand-held devices has led to more frequent reports of weak tornadoes that produce minimal damage. In contrast, strong to violent tornadoes have been observed more consistently over time. Note that tornado strength is measured by the damage it produces, not by the visual appearance of the funnel. Limited real-time observational capabilities in earlier decades did not prevent identification because strong to violent tornadoes leave much more damage which will be assessed later even if the tornado itself was not observed. Since statistics began in 1950, there has been a substantial decrease (by about 50%) in the number of strong to violent tornadoes as shown in Fig. 6.5.1a.

To summarize, there is a noticeable downward trend in the number of severe tornadoes in the U.S. since 1950. After 1990 the number of weak tornadoes in the U.S. has remained roughly constant; data before that are incomplete due to limited monitoring.

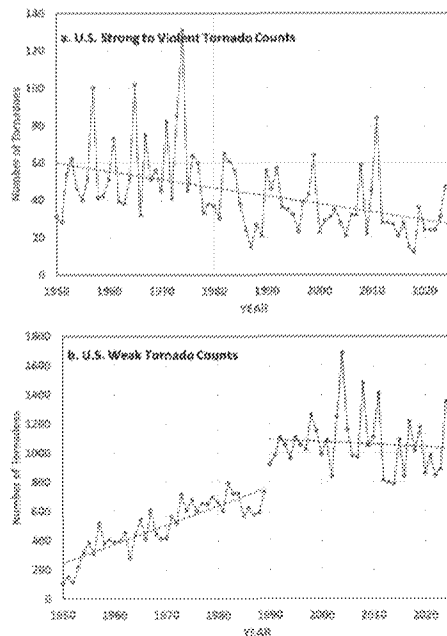


Figure 6.5.1. Annual U.S. tornado counts for (a) strong to violent tornadoes (EF3 to EF5), and (b) weak tornadoes (EF0 to EF2). Based upon NOAA Storms

Prediction Center data, available at https://www.spc.noaa.gov/wcm/data/1950-2024_actual_tomadoes.csv

6.6 Flooding

Changes in floods were assessed as follows:

AR6: The SREX **assessed low confidence for observed changes in the magnitude or frequency of floods** at the global scale. This assessment was confirmed by the AR5 report. The SR15 found increases in flood frequency and extreme streamflow in some regions, but decreases in other regions. . . [H]ydrological literature on observed flood changes is heterogeneous, focusing at regional and sub-regional basin scales, making it difficult to synthesise at the global and sometimes regional scales. (Chapter 11.5)

AR6: [T]he seasonality of floods has changed in cold regions where snowmelt dominates the flow regime in response to warming (*high confidence*). Confidence about peak flow trends over past decades on the global scale is *low*. (Chapter 11.5)

NCA4: Trends in extreme high values of streamflow are **mixed** across the United States. Analysis of 200 U.S. stream gauges indicates areas of both increasing and decreasing flooding magnitude **but does not provide robust evidence** that these trends are attributable to human influences (pp. 240-241)

The absence of detectable US-wide trends in flooding is consistent with the findings in Section 6.4 of absence of coherent changes in extreme precipitation.

6.7 Droughts

Assessments of drought trends were as follows.

AR6: Few AR6 regions show observed increases in meteorological drought (Section 11.9, p. 1575),

AR6: Increasing trends in agricultural and ecological droughts have been observed on all continents (*medium confidence*), but decreases only in one AR6 region (*medium confidence*). Increasing trends in hydrological droughts have been observed in a few AR6 regions. (Chapter 11 Summary)

NCA4: As a consequence of this increased precipitation, **drought statistics over the entire CONUS have declined**. (p. 233)

NCA4: Recent droughts and associated heat waves have reached record intensity in some regions of the United States; however, by geographical scale and duration, the Dust Bowl era of the **1930s remains the benchmark** drought and extreme heat event in the historical record (very high confidence). (p. 231)

SREX: From a paleoclimate perspective, **recent droughts are not unprecedented**, with severe ‘megadroughts’ reported in the paleoclimatic record for Europe, North America, and Australia. (p. 170)

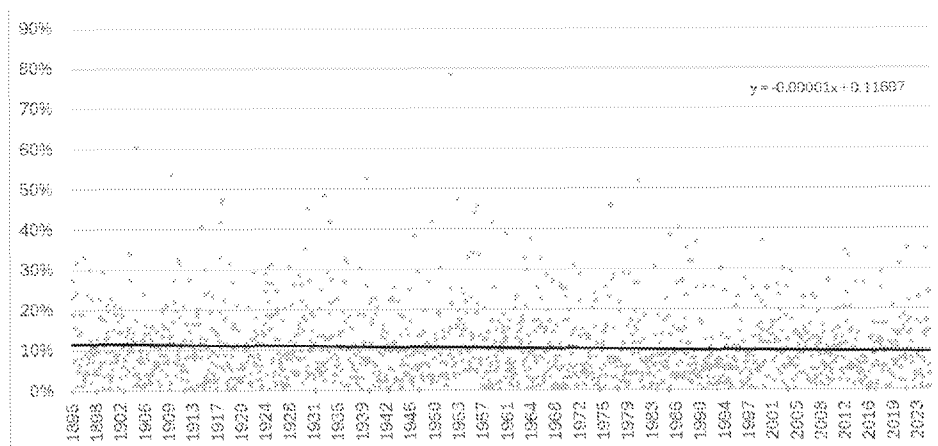


Figure 6.7.1: Monthly percent of US classified as “Very Dry” 1895—2025. Data source: NOAA [[HYPERLINK "https://www.ncsl.noaa.gov/access/monitoring/uspa/wet-dry/0/"](https://www.ncsl.noaa.gov/access/monitoring/uspa/wet-dry/0/)] least squares trendline added.

As shown in Figure 6.7.1, U.S. long-term data shows an insignificantly declining trend in extreme dryness (-0.001 percent per year)

Kogan *et al.* (2020) examines a 38-year high-resolution satellite-based drought measure and concludes that global drought has not intensified and is not connected to climate change: “it is possible to state firmly that global and main grain countries’ drought area and intensity trends have not been following global climate warming since 1980’s.”

In summary there is no evidence of increasing drought frequency or intensity in the U.S. or globally over recent decades.

6.8 Wildfires

The IPCC has not provided an assessment of wildfires. As shown in Figure 6.8.1, global wildfire activity as measured by European Space Agency shows a downward trend in the 21st century.

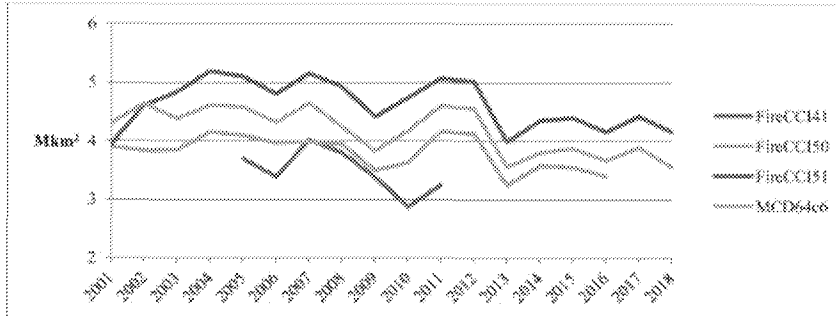


Figure 6.8.1: Global wildfire area 2001-2018. Source: From Lizundia-Loiola *et al.* (2021) Figure 12. The different coloured lines represent data products derived from different satellites and algorithms

Global data show that wildfire coverage is constant or declining on every continent (Samborska and Ritchie, 2024).

Active fire suppression since 1900 makes it difficult to establish a natural baseline for wildfire activity in the U.S. Paleoclimatic evidence indicates that past activity was much higher than today. Marlon *et al.* (2012) used sedimentary charcoal layers to reconstruct fire history of western U.S. for the past 1400 years and also fit a model to predict fire activity as a function of climatic conditions. Their results are summarized in Figure 6.8.2 below (from Figure 2 in their paper). There has been a growing wildfire deficit over the 20th century. In other words, however much fire was observed in the 20th century, it was less than what would have been observed in previous centuries based on the climatic conditions. Parks *et al.* (2025) likewise find that despite the recent increase in wildfire burn area in North America, a significant wildfire deficit remains relative to historical wildfire regimes.

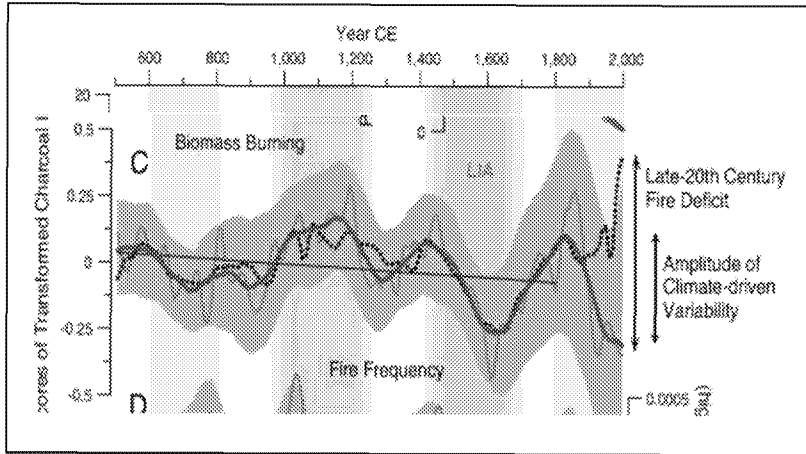


Figure 6.8.2 Fire frequency and fire deficit in the US. The red line shows the smoothed charcoal record and the black dotted line shows the predicted charcoal record based on climatic conditions model. Source: Marlon *et al.* (2012) Figure 2.

U.S. data from the National Interagency Fire Centre (NIFC) from 1926 to 2023 are shown in Figure 6.8.3. The NIFC has removed the pre-1960 data from its current website on the grounds that measurement methods changed after 1960 making the comparison unreliable. Nonetheless just focusing on the post-1985 interval the number of fires is not increasing. The area burned did increase but only until about 2007.

Forest fires have always been a part of nature, and they can certainly create conditions that are inhospitable in the short term for all life, including humans. Science has confirmed the overall benefit and necessity of forest fires. While recent high-profile fires and seasons serve as a reminder of the potential destructive impact, the highest profile U.S. forest fire remains the 1910 Big Blowup fire in the U.S. west, which destroyed over three million acres and eliminated entire towns like Taft, MT (Apple, 2020). The 1910 fire reshaped the U.S. Forest Service (National Forest Foundation 2022) leading to a focus on fire suppression with a primary goal of defeating all forest fires (Forest History Society, 2022). This led to the “10 am rule” in 1935 requiring that all fires spotted on any day had to be controlled by 10 am the following day (National Forest Foundation, 2022).

While defeating all fires seemed a noble goal, questions began to arise as to whether this behavior “followed the science” (U.S. Forest Service, 2022). Over time the U.S. Forest Service has begun to rethink its goals, recognizing that new approaches such as prescribed burns, fuel elimination, and controlled wildfires are more appropriate (Sommer, 2016). Recent research is validating this approach and recognizing that more frequent smaller fires likely result in healthier forests, water ecosystems and biodiversity (Stephens *et al.*, 2021).

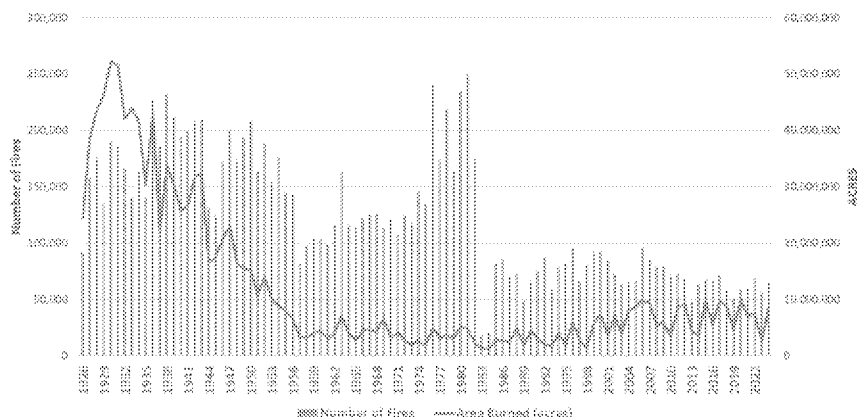


Figure 6.8.3: U.S. wildfires 1926 to 2023. Source: Post-2018: National InterAgency Fire Center data [[HYPERLINK "https://www.nifc.gov/fire-information/statistics/wildfires"](https://www.nifc.gov/fire-information/statistics/wildfires)]. Pre-2017 webarchive.org (n.d.).

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7 CHANGES IN SEA LEVEL

Chapter Summary

Since 1900, global average sea level has risen by about 8 inches. Sea level change along U.S. coasts is highly variable, associated with local variations in processes that contribute to sinking and also with ocean circulation patterns. The largest sea level increases along U.S. coasts are Galveston, New Orleans, and the Chesapeake Bay regions - each of these locations is associated with substantial local land sinking (subsidence) unrelated to climate change.

Extreme projections of global sea level rise are associated with an implausible extreme emissions scenario and inclusion of poorly understood processes associated with hypothetical ice sheet instabilities. In evaluating the IPCC AR6 projections to 2050 (with reference to the baseline period 1995–2014), almost half of the time interval has elapsed by 2025, with sea level rising at a lower rate than predicted. U.S. tide gauge measurements reveal no obvious acceleration beyond the historical average rate of sea level rise.

7.1 Global sea level rise

Global sea level rise is arguably the most important climate impact driver that is unambiguously associated with increasing temperatures. At the global level, warming raises sea level through thermal expansion of sea water and through melting of glaciers and ice sheets. Variations in land water storage are another important factor. At the regional scale, sea level change is influenced by large-scale ocean circulation patterns, and geologic processes and deformation from the redistribution of ice and water. Locally, vertical land motion from geologic processes, ground water withdrawal, and fossil fuel extraction are also important.

The IPCC AR6 estimates that global mean sea level increased by 7.9 (5.9–9.8) inches between 1901 and 2018, with the rate of sea level rise accelerating in recent decades. At the ocean basin scale, sea levels have risen fastest in the Western Pacific and slowest in the Eastern Pacific over the period 1993–2018 (Fox-Kemper *et al.*, 2021). The rate of global sea level rise is estimated to be 0.12 inches/year, about the height of two stacked pennies (NASA, 2020).

The observing systems for global sea level rise have advanced significantly in the satellite era, particularly with the advent of satellite altimeters in 1993. Local tide gauges have provided useful data for the past century, and even longer for a few locations. Following the end of the Little Ice Age in the mid-nineteenth century, tide gauges show that the global mean sea level began rising during the period 1820–1860, well before most anthropogenic greenhouse gas emissions.

7.2 U.S. sea level rise

Observed and predicted rates of mean global sea level rise might have little scientific relevance for specific locations, owing to local processes (NOAA, 2025). Figure 7.1 shows that in Canada and Alaska (and also northern Washington), sea level is decreasing, owing to uplift from glacial rebound. Most of the Pacific coast tide gauges show low rates of sea level rise, while largest U.S. rates are on the Gulf coast (Louisiana and Texas) and in the mid-Atlantic states (Chesapeake Bay region).

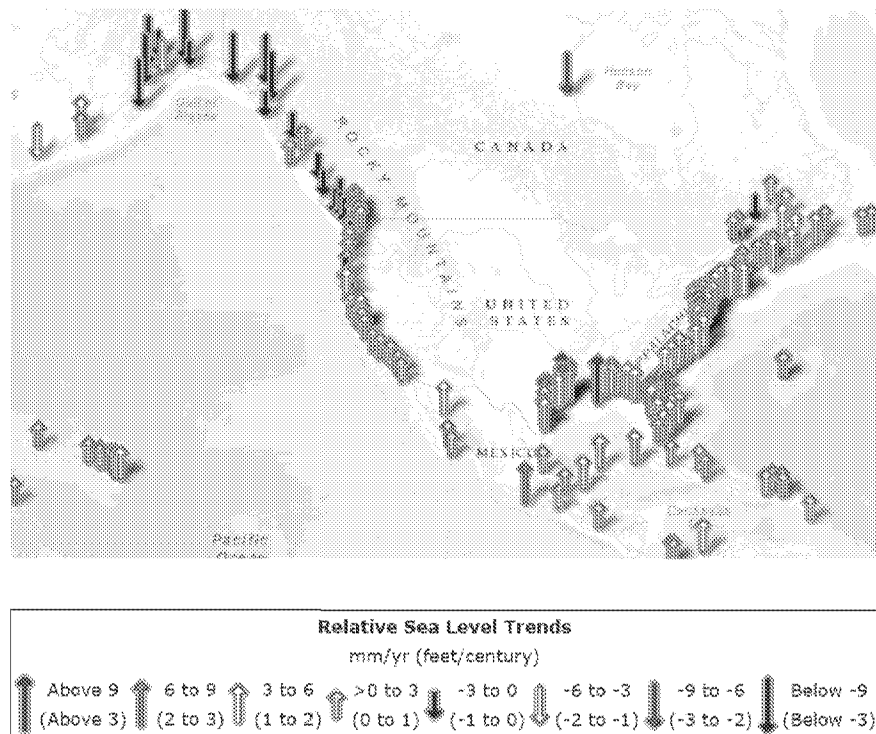


Figure 7.1. Map of rates of relative sea level rise along the U.S. coast (NOAA, <https://tidesandcurrents.noaa.gov/strends/>). For reference, 3 mm = 0.12 in.

Measurements of relative sea level rise from tide gauges conflate the climatologically relevant increase in the volume of seawater with local vertical land motion. The latter, which varies from place to place, is best measured using a Global Positioning System (GPS) station located near the tide gauge. It is driven by a range of processes that can be comparable to the climate signal and can locally exacerbate (subsidence/sinking) or mitigate (uplift) the risk of sea level rise (Wöppelmann and Marcos, 2016). Human activities triggering subsidence are often significant. They include soil drainage (*e.g.* for urban development) and subsurface extraction of groundwater or hydrocarbons.

Table 7.1 shows absolute sea-level rise (ASLR) for selected locations, determined from the sum of uncorrected sea-level rise as estimated from tide gauge time series (RSLR) (NOAA, 2025) and the vertical land motion (VLM) measurements (NAS, 2012; Letetrel *et al.*, 2015; Karegar *et al.*, 2016). The absolute sea level rise for each of these locations is significantly smaller than the measured relative sea level rise owing to local subsidence. More than half of the measured relative sea level rise is attributed to land sinking for these locations: San Francisco, Galveston, Grand Isle. For reference, the global average rate of absolute sea level rise is estimated to be 0.12 inches/year.

Location	RSLR	VLM	ASLR
San Francisco, CA	+0.08	-0.06	+0.02
Galveston, TX	+0.26	-0.19	+0.07
Grand Isle, LA	+0.36	-0.28	+0.08
St Petersburg, FL	+0.12	-0.02	+0.10
New York City, NY	+0.11	-0.05	+0.06

Table 7.1 Absolute sea level rise (inches/year) consisting of Relative Sea Level Rise (RSLR) plus Vertical Land Motion (VLM)

San Francisco Bay

Over the past 100 years, relative sea level in the San Francisco Bay area has risen by 7.8 inches, at an average rate of 0.08 inches/year (Figure 7.2). As shown in Table 7.1, San Francisco's vertical land motion is -0.06 inches/year (sinking), producing a recent absolute rate of +0.02 inches/year. Portions of Treasure Island, San Francisco International Airport, and Foster City are sinking as fast as 0.4 inches/year (Shirzaei and Bürgmann, 2018). Problems in the San Francisco Bay area, including threats to the airport, are caused primarily by soil compaction in landfill zones that were formerly wetlands, not by the slow creep of global sea level rise.

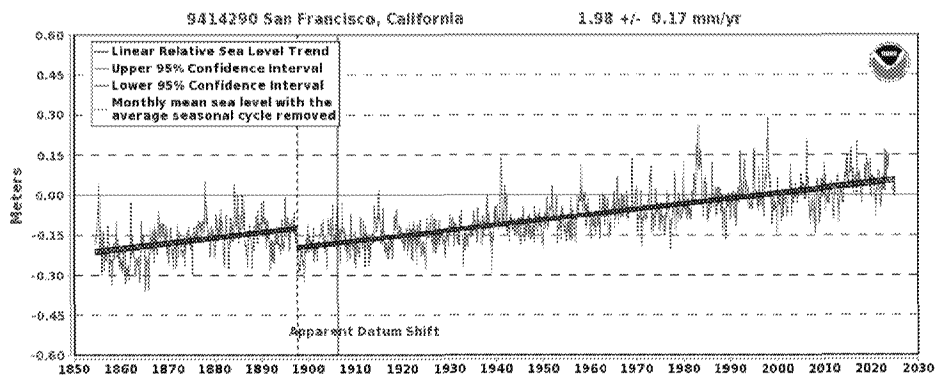


Figure 7.2. Tide gauge measurements at San Francisco, California, obtained from NOAA - https://tidesandcurrents.noaa.gov/strends/strends_station.shtml?id=9414290 (downloaded 4/22/25).

Galveston - Houston

Long-term tide measurements at Galveston Pier 21 show sea level rise of 2.18 feet in the past century, or a rate of 0.26 inches/year (Figure 7.3). Vertical land motion (subsidence) at Galveston is estimated at -0.19 inches/year, yielding an absolute rate of rise of +0.07 inches/year (Table 7.1). The U.S. Geologic Survey found that most of the land-surface subsidence in the Houston-Galveston region is a direct result of groundwater withdrawals (Kasmarek and Ramage 2017), which caused compaction of the aquifer sediments, mostly in the fine-grained silt and clay layers. By 1979, as much as 10 feet of subsidence had occurred in Houston.

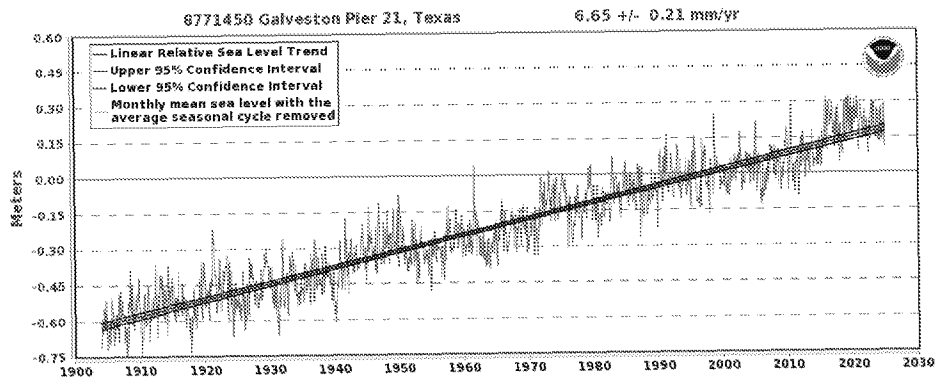


Figure 7.3. Tide gauge measurements Galveston Pier, TX, obtained from NOAA - https://tidesandcurrents.noaa.gov/sltrends/sltrends_station.shtml?id=8771450 (downloaded 4/22/2025).

New Orleans and the Mississippi delta

Long-term tide gauge measurements at Grand Isle, Louisiana, show that sea level has risen by slightly more than 3 feet over the last 100 years at an average rate of 0.36 inches/year (Figure 7.4). Vertical land motion (subsidence) is estimated -0.28 inches/year. Table 7.1 gives the absolute sea level rise as +0.08 inches/year.

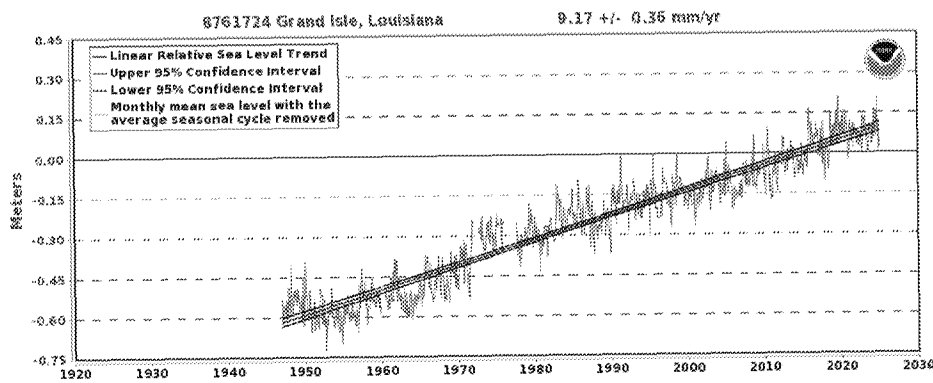


Figure 7.4. Tide gauge measurements at Grand Isle, LA, obtained from NOAA - https://tidesandcurrents.noaa.gov/sltrends/sltrends_station.shtml?id=8761724 (downloaded 4/22/25).

The issues of sea level rise and land loss in the Mississippi Delta region are complex, with geological subsidence and the decline in sediment transported by the Mississippi River being the dominant drivers. The construction of dams in the basin since the 1950s has decreased the Mississippi's suspended sediment load by ~50 percent (Maloney 2018). A new subsidence map of coastal Louisiana finds the coastal region to be sinking at about one third of an inch per year (Neinhuis *et al.* 2017), associated with groundwater and resource withdrawal. As the city's elevation averages one to two feet *below* sea level, sea level rise from anthropogenic warming is hardly the dominant driver of New Orleans' problems.

New York City

New York City is particularly vulnerable to the effects of sea level rise because it is built primarily on islands and has 520 miles of coastline. Measurements by a tide gauge at the southern tip of Manhattan (The Battery) show that relative sea level has risen over 11 inches over the past century, at an average rate of 0.11 inches/year (Figure 7.5). But vertical land motion in the New York City area is -0.05 inches/year (roughly 5 inches per century), so that the absolute rate of sea level rise at The Battery is 0.06 inches/year, or about 55 percent of the measured relative sea level rise.

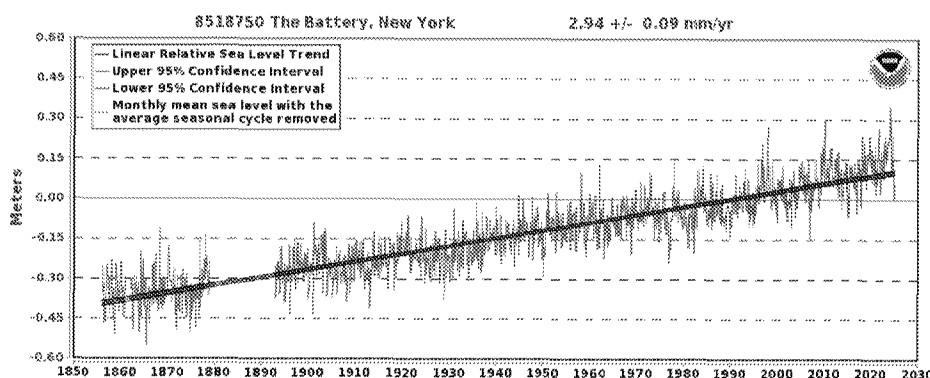


Figure 7.5. Tide gauge measurements at The Battery, New York, obtained from NOAA - https://tidesandcurrents.noaa.gov/strends/strends_station.shtml?id=8518750 (downloaded 4/22/25).

7.3 Projected sea level rise

The concern over sea level rise is not about the roughly eight inches of global rise since 1900. Rather, it is about projections of *accelerated* rise based upon simulations of a warming climate through the 21st century.

AR6 finds *high agreement* across published global mean sea level projections for 2050 with little sensitivity to emissions scenarios. Considering only projections incorporating ice-sheet processes in whose quantification there is at least *medium confidence*, the global sea level projections for 2050, across all emissions scenarios, fall between 3.94 and 15.75 inches (5th–95th percentile *very likely* range) relative to the 1995–2014 baseline period (Fox-Kemper *et al.*, 2021).

Conversely, AR6 states there is *low agreement* across published global mean sea level projections for 2100, particularly for higher emissions scenarios. Considering only projections representing ice sheet processes in whose quantification there is at least *medium confidence*, the AR6 global mean sea level projections for 2100 lie between 7.87 and 39.37 inches (5th–95th percentile *very likely* range) under the medium emissions scenario SSP2–4.5 (Fox-Kemper *et al.*, 2021). There is deep uncertainty surrounding projections of sea level rise to 2100 owing to uncertainties in ice sheet instabilities, particularly for the higher emissions scenarios.

In February 2022, NOAA issued its projections of sea level rise for various sites along the U.S. coast (Sweet *et al.*, 2022). They claim that by 2050, the sea will have risen one foot at The Battery in Manhattan (relative to 2020). A one-foot rise in thirty years would be more than twice the current rate and about three times the average rate over the past century. In that historical context, NOAA's projection is remarkable—as shown in Figure 7.6, it would require a dramatic acceleration beyond anything observed since the early 20th century. But even more noteworthy is that Sweet *et al.* (2022) say this rise is “locked in”—it will happen no matter what future emissions are. We should know in a decade or so whether that prediction has legs.

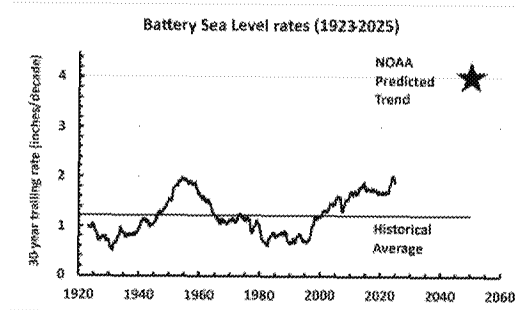


Figure 7.6 Rate of sea level rise at the Battery in Manhattan. Shown is the historical thirty-year trailing trend, together with the allegedly “locked in” NOAA predicted trend for 2050. Historical data: NOAA Tides and Current.

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8 UNCERTAINTIES IN CLIMATE CHANGE ATTRIBUTION

Chapter summary

“Attribution” refers to identifying the cause of some aspect of climate change, specifically with reference to anthropogenic activity. There is an ongoing scientific debate around attribution methods, particularly regarding extreme weather events. Attribution is made difficult by high natural variability, the relatively small expected anthropogenic signal, lack of high-quality data, and reliance on deficient climate models. The IPCC has long cautioned that methods to establish causality in climate science are inherently uncertain and ultimately depend on expert judgement.

Substantive criticism of the main IPCC assessments of the role of CO₂ in recent warming focus on inadequate assessment of natural climate variability, uncertainties in measurement of solar variability and in aerosol forcing, and problems in the statistical methods used for attribution.

The IPCC does not make attribution claims for most climate impact drivers related to extreme events. Statements related to statistics of global extremes (*e.g.* event probability or return times, magnitude and frequency) are not generally considered accurate owing to data limitations and are made with low confidence. Attribution of individual extreme weather events is challenging due to their rarity. Conflicting claims about the causes of the 2021 Western North America Heatwave illustrate the perils of hasty attribution claims about individual extreme events.

8.1 Introduction

The Intergovernmental Panel on Climate Change (IPCC) distinguishes between detection of climate change and attribution of its cause. As defined by the AR6 WGI Glossary (IPCC, 2025):

Detection: Detection of change is defined as the process of demonstrating that climate or a system affected by climate has changed in some defined statistical sense, without providing a reason for that change. An identified change is detected in observations if its likelihood of occurrence by chance due to internal variability alone is determined to be small, for example, <10%.

Attribution: Attribution is defined as the process of evaluating the relative contributions of multiple causal factors to a change or event with a formal assessment of confidence.

Both detection and attribution rely on statistical analysis. Detection focuses on whether changes are significant enough to stand out from random variation regardless of cause. Attribution involves comparison of observed events to model-generated counterfactuals. Since experimentation on the climate is not possible, attribution requires statistical inference to assess how much human activities (such as GHG emissions) versus natural factors (like volcanic eruptions) contribute to observed changes. Attribution methods assume all external and internal drivers of the system are known and represented.

There are ongoing scientific debates around attribution methods, especially those for attributing extreme weather events to climate change. The IPCC has long cautioned that methods to establish causality in climate science are inherently uncertain and ultimately depend on expert judgement. AR4 Working Group I (Hegerl *et al.*, 2007) explained it as follows:

‘Attribution’ of causes of climate change is the process of establishing the most likely causes for the detected change with some defined level of confidence... unequivocal attribution would require controlled experimentation with the climate system. Since that is not possible, in practice attribution of anthropogenic climate change is understood to mean demonstration that a detected change is ‘consistent with the estimated responses to the given combination of anthropogenic and natural forcing’ and ‘not consistent with alternative, physically plausible explanations of recent climate change that exclude important elements of the given combination of forcings’ (IPCC, 2001)... The approaches used in detection and attribution research described above cannot fully account for all uncertainties, and thus ultimately expert judgement is required to give a calibrated assessment of whether a specific cause is responsible for a given climate change.

AR5 Working Group II (Cramer *et al.*, 2014) makes the following statement:

Two broad challenges to the detection and attribution of climate change impacts relate to observations and process understanding. On the observational side, high-quality, long-term data relating to natural and human systems and the multiple factors affecting them are rare. In addition, the detection and attribution of climate change impacts requires an understanding of the processes by which climate change, in conjunction with other factors, may affect the system in question.

Because of the complexity of the causal chains in the climate system, investigation of these relationships is exceptionally challenging.

8.2 Attribution methods

The IPCC employs several attribution methods to assess the causes of observed climate changes, distinguishing between natural and human-induced factors. Below is a concise description of the key IPCC attribution methods.

Optimal Fingerprinting uses linear regression to explain variations in observed climate data as a weighted sum of climate model simulations run with and without anthropogenic forcings. The data used in the regression model is weighted to try to minimize the influence of random noise and the estimation method is chosen to account for climate model error.

Time Series Analysis exploits statistical differences between anthropogenic forcing and natural variability to see which dominates observed temperatures and also uses variations in the timing of changes to determine if causal dependence across data series can be inferred.

Process-Based Attribution focuses on understanding the physical mechanisms behind observed changes. This approach combines observations, climate models, and theoretical understanding to attribute changes in specific processes to forcings. This approach is often used for regional climate phenomena, such as monsoon changes or polar amplification

Extreme Event Attribution assesses the role of human influence in the likelihood of occurrence or intensity of extreme weather events (*e.g.* heat waves or droughts). Methods include:

- Probabilistic Event Attribution uses large ensembles of climate model simulations to compare observed events to model-generated counterfactuals
- Storyline Approach examines the physical processes driving an event and evaluates how anthropogenic forcings might have modified those processes

8.3 Attribution of global warming

Attribution statements for global warming in the three most recent IPCC Assessment reports are:

AR4: **Most of the observed increase** in global average temperatures since the mid-20th century is *very likely* due to the observed increase in anthropogenic greenhouse gas concentrations. (IPCC 2007)

AR5: It is *extremely likely* that **more than half of the observed increase** in global average surface temperature from 1951 to 2010 was caused by the anthropogenic increase in greenhouse gas concentrations and other anthropogenic forcings together. The best estimate of the human-induced contribution to warming is similar to the observed warming over this period. (IPCC 2013)

AR6: The *likely* range of total human-caused global surface temperature increase from 1850–1900 to 2010–2019 is 0.8°C to 1.3°C, with a best estimate of 1.07°C. It is *likely* that well-mixed GHGs contributed a warming of 1.0°C to 2.0°C, other human drivers (principally aerosols) contributed a cooling of 0.0°C to 0.8°C, natural drivers changed global surface temperature by –0.1°C to +0.1°C, and internal variability changed it by –0.2°C to +0.2°C. It is *very likely* that **well-mixed GHGs were the main driver** of tropospheric warming since 1979. (IPCC 2021)

The AR4 and AR5 attribution statements reference the warming since the mid-20th century, the period when greenhouse gas emissions began increasing rapidly. The words “most” and “more than half” are deliberately imprecise, potentially ranging from 51 to 99 percent of the warming – presumably this imprecision is to account for structural uncertainties such as natural internal variability. The confidence level increases from *very likely* to *extremely likely* from AR4 to AR5. The structure of the attribution statement in the AR6 is fundamentally different, referencing the warming to the period 1850–1900. The AR6 attribution statement is more precise numerically, but with a lower level of confidence at “likely” – AR6 attributes essentially all the warming to increases in greenhouse gases. The most confident statement from AR6 relates to the tropospheric warming since 1979, using the words “main driver” and “*very likely*.”

There are three areas of substantive criticism of the IPCC’s assessment of the causes of the recent warming: inadequate assessment of natural climate variability, inappropriate statistical methods, and substantial discrepancies between models and observations. The last is discussed in Chapter 5, while this chapter discusses the first two factors. All of these criticisms are relevant to the IPCC’s policy-informing attribution of the recent warming, which also underpins extreme event attribution

8.3.1 Natural climate variability

AR6 states that natural external drivers since 1850–1900 have changed global surface temperature by –0.1°C to +0.1°C, and internal variability has changed it by –0.2°C to +0.2°C – on average having essentially no net impact on the warming since 1850–1900. As discussed below, this minimal contribution of natural variability has been disputed by several publications that question the magnitudes of solar variability and internal variability from large-scale ocean circulations.

Solar variability

AR5 concluded that the best estimate of radiative forcing due to Total Solar Irradiance (TSI) changes over the period 1750–2011 was very small (0.05 W/m², Myrhe *et al.*, 2014). AR6 acknowledges substantially higher values and a much larger range of estimates of changes in TSI over the last several centuries, stating that the TSI between the Maunder Minimum (1645–1715) and the second half of the twentieth century increased by 0.7–2.7 W/m², a range that includes both low and high variability TSI data

sets (Gulev, 2021). However, the recommended forcing dataset for the CMIP6 climate model simulations used in AR6 for attribution studies averages two data sets with low solar variability (Matthes, 2017).

While AR6 shows a substantially greater solar impact than does AR5, the overall impact of solar forcing on the climate was still assessed to be small compared to anthropogenic forcing. However, the impact of solar variations on the climate is uncertain and subject to substantial debate (Lockwood, 2012; Connolly *et al.*, 2021) - something that is not evident in the IPCC assessment reports.

The variations of TSI over time remains a challenging problem. Since 1978, there have been direct measurements of TSI from satellites. However, the data exhibits non-negligible inconsistencies, and interpreting any multi-decadal trends in TSI requires comparisons of observations from overlapping satellites. There are several rival composite TSI datasets that disagree as to whether TSI increased or decreased during the period 1986–96 (the ACRIM gap; see Chapter 4). Further, the satellite record of TSI is used to calibrate proxy models that infer past solar variations from sunspots and cosmogenic isotope measurements (Velasco Herrera *et al.*, 2015).

There is substantial evidence for high solar activity in the second half of the 20th century (starting in 1959) and extending into the 1990's, before a decline in the early 21st century; this period is often termed the “Modern Maximum.” (Chatzistergos *et al.*, 2023; Solanki *et al.*, 2004; Usoskin *et al.*, 2007). However, some scientists have concluded that it is not possible to be confident of any multi-decadal trend in TSI (Schmutz, 2021).

This uncertainty causes some reconstructions of TSI from 1750 to have low variability (implying a very low impact of solar variations on global mean surface temperature) whereas datasets with high TSI variability can explain more than 70 percent of the temperature variability since preindustrial times (Scafetta, 2013; Stefani, 2021). The choice of TSI satellite record used in an analysis can therefore substantially influence how much climate change is attributed to human versus natural forcings.

There is growing evidence that other aspects of solar variability, which are referred to as solar indirect effects, either amplify TSI forcing or are independent of TSI forcing. Scafetta *et al.* (2023) suggests that ~80 percent of solar influence on climate might stem from non-TSI mechanisms. There are numerous candidate processes, including solar ultraviolet changes; energetic particle precipitation; atmospheric-electric-field effect on cloud cover; cloud changes produced by solar-modulated galactic cosmic rays; large relative changes in the magnetic field; and the strength of the solar wind. Such solar indirect effects are not included in climate models, although indirect methods of estimating their impacts suggest they are significant. However, claims of non-TSI mechanisms influencing climate are uncertain and debated.

Natural variability of large-scale ocean circulations

Variations in global mean surface temperature are linked to recurrent large-scale variations in ocean circulation patterns, including the Atlantic Multidecadal Oscillation (AMO), the Pacific Decadal Oscillation (PDO) and the El Niño-Southern Oscillation (ENSO). These circulations influence ocean heat uptake and heat distribution and also influence atmospheric circulation patterns and cloud distributions. There is some debate as to whether these variations are strictly internal to the climate system, or whether this variability can have a solar/astronomical origin or can be influenced by large volcanic eruptions.

While climate models simulate the large-scale ocean circulations and internal climate variability, most models have too little amplitude compared to observations at multi-decadal frequencies and phasing out-of-sync with the observed climate (Kravtsov *et al.* 2024). Averaging multiple simulations effectively averages out the internal variations, leaving only the forced climate variability (*e.g.* CO₂ forcing). With most of the modern warming beginning in the late 1970's, the recent 50-year warming is on the same time scale as the multi-decadal oscillations of the AMO and PDO.

Here are summary statements from the AR5 and AR6 reports:

AR5: Decadal variability in the Pacific, associated with the PDO or IPO [Interdecadal Pacific Oscillation], contributes significantly to regional and global temperature trends, but the relative contributions of internal variability and external forcing are difficult to disentangle in CMIP5 simulations.

AR6: Since AR5, there has been increased understanding of the role of internal variability, such as ENSO, PDO, and AMO, in modulating regional climate trends. However, limitations in simulating the exact timing and amplitude of these modes in CMIP6 models contribute to uncertainties in attributing observed changes to anthropogenic forcing (*high confidence*).

The amplitude of the peak-to-trough impact of the multi-decadal oscillations on global temperatures has been assessed by the IPCC AR6: “The likely range of change due to internal variability is -0.2°C to $+0.2^{\circ}\text{C}$ (IPCC, 2021).” This implies a trough-to-peak change of 0.4°C . Over many centuries, any global temperature changes from the troughs and peaks will cancel out, with little to no net impact.

However, with a nominal timescale of 60-80 years for oscillations such as the AMO and PDO, the timing of the peaks and troughs can be confused with the secular trend. This becomes very relevant for attributing the warming for the past 50 years, when most of the recent warming has occurred. Even if climate models have the correct amplitude of the multidecadal oscillations, the timing of the peaks and troughs is not adequately simulated, with each model and individual ensemble simulating a different phasing. Once the simulations of the ensemble members and different models are averaged, the natural internal variability contribution is averaged out owing to differences in phasing, effectively cancelling the role of natural internal variability in the attribution process.

The time series of global surface temperature anomalies since 1850 (Figure 8.1) shows irregular variations of significant amplitude against the background of an overall warming trend and, after 1976, a pronounced difference in trends between the Northern Hemisphere and the Southern Hemisphere. The period between 1905-1945 shows a strong trend of warming. The following 30 years from 1945-1976 showed slightly declining temperatures. The most recent warming period started in 1977.

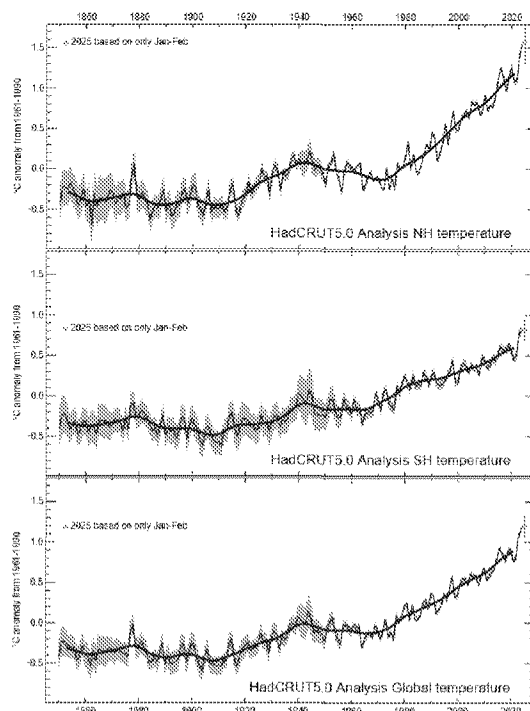


Figure 8.1 Global average surface temperature anomalies 1850—2025. Top: Northern hemisphere. Middle: Southern hemisphere. Bottom: Global average. Source: UK Hadley Centre [[HYPERLINK "https://crudata.uea.ac.uk/cru/data/temperature/HadCRUT5.0Analysis.pdf"](https://crudata.uea.ac.uk/cru/data/temperature/HadCRUT5.0Analysis.pdf)]

The causes of the early 20th century warming are discussed by Hegerl *et al.* (2017; 2019). The atmospheric CO₂ concentration increased from 298 ppm in 1905 to 310 ppm in 1941, implying that CO₂ had little impact. Volcanic activity during this period was very low, and solar forcing is uncertain. Yet Hegerl *et al.* (2017) somehow inferred that 40–54 percent of this warming could be attributed to external forcing, with the rest associated with internal variability. Bronniman *et al.* (2024) focused on the causes of the cooling in the first decade of the 20th century in the Southern Hemisphere. They found that the cooling was related to a La-Niña-like pattern in the Pacific, a cold tropical and subtropical South Atlantic, a cold extratropical South Pacific, and cool southern midlatitude land areas. The Southern Annular Mode was positive, with a strengthened Amundsen–Bellingshausen seas low, although the spread of the data products is considerable.

The warming in the 1930's and subsequent cooling during mid-century was particularly pronounced in the Arctic. Bokuchava and Semenov (2021) find that these variations were most likely caused by a combined effect of long-term natural climate variations in the North Atlantic and North Pacific with a contribution of the natural radiative forcing related to the reduced volcanic activity and variations of solar activity, as well as growing greenhouse gases concentrations. Tokinaga *et al.* (2017) showed that the combined effect of internally generated Pacific and Atlantic interdecadal variability intensified Arctic

warming in the early 20th century. The synchronized Pacific-Atlantic warming drastically alters planetary-scale circulations over the Northern Hemisphere; these same circulation patterns have a global influence.

The cooling period between 1945 and 1976 has been referred to as the “grand hiatus.” Numerous causes have been hypothesized: natural internal variability from fluctuations associated with the Pacific Decadal Oscillation and Atlantic Multidecadal Oscillation; cooling from increased emissions of aerosols from industrial activities; increased heat uptake in the Atlantic, Southern and Equatorial Pacific Oceans. Thompson *et al.* (2010) find that the hemispheric differences in temperature trends in the middle of the twentieth century stem largely from a rapid drop in Northern Hemisphere sea surface temperatures of about 0.3 °C between about 1968 and 1972.

The Great Pacific Climate Shift of 1976-1977 was a notable climatic event characterized by an abrupt change in the North Pacific Ocean's atmosphere-ocean system that interacted with global climate patterns. This shift is closely associated with the Pacific Decadal Oscillation (PDO) that oscillates between warm and cool phases over decades. The 1976 shift marked a transition from a predominantly negative (cool) PDO phase (1947–1976) to a positive (warm) phase (1977 through 2000), with significant implications for global and regional climate patterns, including the frequencies of El Niño and La Niña events. The Great Pacific Climate Shift coincided with the beginning of a period of accelerated global warming. When the Great Pacific Climate Shift is accounted for in climate attribution analyses since 1950, 40 percent or more of the warming in the second half of the 20th century is attributed to natural internal variability (McLean *et al.*, 2009; Tung and Zhou, 2013; Chylek *et al.*, 2016; Scafetta, 2021).

8.3.2 Optimal fingerprinting

Optimal fingerprinting is a statistical technique introduced by Allen and Tett (1999) that compares observed climate data to climate model simulations to identify patterns (or “fingerprints”) associated with human or natural forcing. It involves taking a vector of observed climate changes, such as warming rates in locations around the world, and decomposing them into a weighted sum of “signals”, which are analogues to the observations generated by climate models with different types of forcings. The weights are chosen using a regression method called Total Least Squares (TLS). The analysis commonly uses just two signals, one generated by models using only anthropogenic forcings and one using only natural forcings. If the estimated weighting coefficient attached to a signal is significantly different from zero, then that signal is said to be “detected”. If it is close to 1.0, then the model generating the signal is said to be consistent with observations. If it is less than 1.0, then the model signal for that forcing is too strong and needs to be scaled down, and vice-versa. Both the observed data and the signals are weighted using a model-generated estimate of patterns of randomness in the climate system so as, in principle, to put maximum weight on regions where natural variability is minimized.

Optimal fingerprinting is the primary tool for attribution in the research literature. While it has been widely used and prominently featured by the IPCC since 2001, there is very little literature examining the statistical properties of the results it generates. One of its inherent weaknesses is that results depend on assumptions about the accuracy of climate models, especially regarding their representation of natural variability (IPCC AR5 Ch 10.2.3).

A series of papers by McKittrick (McKittrick 2021, 2022, 2023, 2025) has challenged the optimal fingerprinting method, arguing that it is inherently unreliable and that practices adopted from econometrics can provide more valid results. Statistical theory requires that for regression models to yield unbiased coefficients, a set of assumptions called the Gauss-Markov conditions must hold. McKittrick (2021) argued that the Allen and Tett (1999) methodology violates the Gauss-Markov conditions, leading to potentially biased and inconsistent fingerprinting coefficients, undermining the reliability of the method. Chen *et al.* (2023) confirmed this analysis, although they argued that the method could yield valid results under very stringent assumptions. McKittrick (2022) and (2023) raised a further concern that climate scientists—virtually alone among

scientific disciplines—have used TLS to estimate anthropogenic greenhouse gas signal coefficients, despite its tendency to be unstable unless some strong assumptions hold that in practice are unlikely to be true. Under conditions that easily arise in optimal fingerprinting, TLS estimates have a large positive bias. Thus, any study that used TLS for optimal fingerprinting without verifying its applicability in the specific data context has likely overstated the attribution.

McKittrick (2025) presented an empirical example comparing the results of conventional optimal fingerprinting against methods drawn from mainstream econometrics that are known to be valid for the specific application of signal detection. While the IPCC optimal fingerprinting method yields an anthropogenic signal coefficient close to 1.0 on a global temperature data set spanning 1900 to 2010, consistent estimation yields a coefficient around 0.4, which rises to about 0.65 on data spanning 1980 to 2010, implying the model response to greenhouse gases needs to be scaled down by about half to optimally match observations. The natural forcing signal coefficient, by contrast, is between 2.0 and 4.0, implying the climate model signals of natural forcing need to be scaled up two- to four-fold to match observed climate change. The fingerprinting coefficients estimated in McKittrick (2025), when used to scale the average sensitivity of the climate models used to generate the forcing signals in his data set, imply a Transient Climate Sensitivity of 1.4°C, which is consistent with the estimate by Lewis (2023) using a different estimation method and multiple independent data sets.

While these analyses do not falsify results of the IPCC optimal fingerprinting method, they do challenge the basis on which they were believed to be correct.

8.3.3 Time series methods

The IPCC (AR5 WGI 10.2.2) drew attention to alternative approaches of assessing causality that emerged from the time series econometrics literature. These have the advantage of not depending on assumptions about the accuracy of climate models, but the disadvantage that they depend on difficult-to-test assumptions about the data generating process underlying climate and forcing data. The methods, now referred to as climate econometrics, use the tools of unit root testing, Granger causality and cointegration analysis, all of which are familiar in economics and finance and which are slowly being adopted in climate science. Time series analysis methods hold out the possibility of determining whether anthropogenic or natural forcings are the primary drivers of climate change without requiring the use of climate models, although key questions remain unsettled (e.g. Dergiades *et al.*, 2016, Balcombe *et al.*, 2019, Dagsvik *et al.*, 2020, Razzak, 2022, Dagsvik and Moen, 2023).

“Granger causality” modeling is a tool for determining directions of influence in data series that move together. It is a statistical, not a physical, concept. If knowing the current value of one variable significantly improves the forecast accuracy of another variable, we can infer a causal connection exists; this is called Granger causality. The modeling tools, called vector autoregression, measure direct impulse-response patterns and feedbacks. An application to the well-known Vostok ice core data revealed an error in Al Gore’s documentary *An Inconvenient Truth*. Gore showed the data and drew attention to the coherence of temperature and CO₂ changes over a 440,000 year span, which he asserted was due to CO₂ driving temperature changes. But temperature changes can also affect atmospheric CO₂ levels. Davidson *et al.* (2015) examined the series and found that temperature Granger causes CO₂ but not the reverse. In other words on the time scales represented in the Vostok data, the coherence in the series is primarily due to the influence of temperature on CO₂ levels, not the feedback of CO₂ levels on temperature.

In summary, the primary statistical methods for attributing causation in climate data are optimal fingerprinting and time series analysis. Applications of the Allen and Tett (1999) optimal fingerprinting method dominate the attribution literature and have underpinned past IPCC conclusions, but results depend on the accuracy of climate models and the method has recently been criticized as inherently biased. Time

series methods do not depend on climate models but require assumptions of their own and have generated results that have thus far not converged on a consensus.

8.4 Declining planetary albedo and recent record warmth

A sharp recent increase in global average temperatures has raised the question of short-term drivers of climate. One such candidate is the fraction of absorbed solar radiation which has also increased abruptly in recent years. The question is whether the change is an internal feedback to warming caused by greenhouse gases, or whether something else increased the fraction of absorbed radiation which then caused the recent warming.

The planetary albedo is the fraction of incoming solar radiation that is reflected back into space rather than being absorbed by the planet. Highly reflective surfaces like cloud tops and snow and ice are most important in this regard. The Earth's albedo is approximately 30 percent, meaning almost a third of the sunlight that reaches Earth is directly reflected back to space. A lower albedo implies more solar energy is absorbed by the planet to be then re-radiated as heat. Hence, other things being equal, a decline in planetary albedo is associated with a warming of the Earth.

Arguably the most striking change in the Earth's climate system during the 21st century is a significant reduction in planetary albedo since 2015, which has coincided with at least two years of record global warmth. Figure 8.2 shows the planetary albedo variations since 2000, when there are good satellite observations. The 0.5 percent reduction in planetary albedo since 2015 corresponds to an increase of 1.7 W/m² in absorbed solar radiation averaged over the planet (Hansen and Karecha, 2025). For comparison, the IPCC estimate of forcing from the increase in atmospheric CO₂ compared to preindustrial times is approximately 2.2 W/m² (Figure 3.1.2).

Looking back prior to 2000 with less adequate data, Goessling *et al.* (2025) assessed that the planetary albedo was relatively low around the 1940's and 50's before rising industrial aerosol precursor emissions increased the albedo until the 1980's. The strongest planetary albedo excursions were high-albedo episodes caused by volcanic eruptions, such as after the Mount Pinatubo eruption in 1991. Although uncertain, the 2023 planetary albedo minimum might have been the lowest since at least 1940.

Changes in surface characteristics cannot explain this decrease in planetary albedo since 2015:

- Arctic sea ice extent has declined by about 5 percent since 1980 (https://nsidc.org/data/seaice_index/images/s_plot_hires.png), although following 2007 there has been a pause in the Arctic sea ice decline (England *et al.*, 2025)
- Regarding Antarctic sea ice, the IPCC AR6 concludes that "There has been no significant trend in Antarctic sea ice area from 1979 to 2020 due to regionally opposing trends and large internal variability." (Summary for Policymakers, A.1.5)
- Northern hemispheric annual snow cover has been slowly declining since 1967, with barely significant trends. The data show the Northern Hemisphere has snowier winters, accompanied by more rapid melt in spring and summer, see [[HYPERLINK "http://climate.rutgers.edu/snowcover"](http://climate.rutgers.edu/snowcover)] and Section 5.6
- Global greening (Chapter 2) is contributing to the decrease in planetary albedo, as forests have a lower albedo than open lands or snow. However, there is some evidence that forests increase cloud cover (high reflectivity), which counteracts the direct albedo decrease associated with increasing forested area.

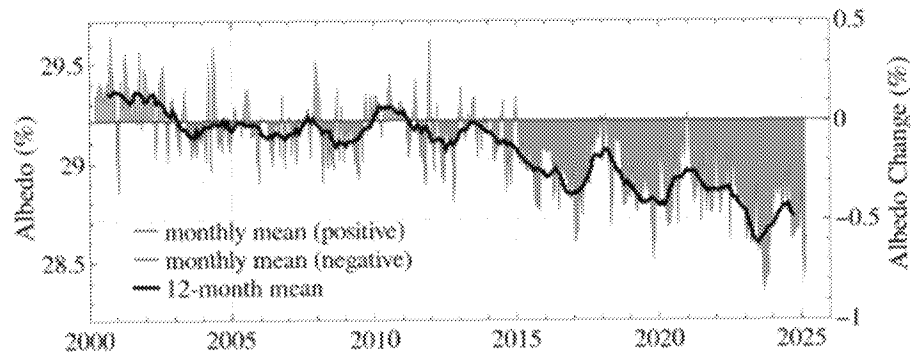


Figure 8.2. Earth's albedo (reflectivity, in percent), with seasonality removed. From Hansen and Karecha (2025)

Changes in surface reflectivity are contributing to a small, slow decline in planetary albedo. However, these changes cannot explain the sharp decline in planetary albedo beginning in 2015. Further, any changes in surface albedo are effectively damped by about a factor of three on average, primarily due to cloud masking (Loeb *et al.* 2019). Surface albedo changes have thus contributed only weakly to the recent planetary albedo decline, particularly when averaged annually and globally. This leaves the most likely explanation for the sharp albedo decline as changes in atmospheric aerosols and/or clouds.

Aerosols are small particles in the atmosphere that reflect sunlight and can interact with cloud processes to make the clouds more reflective. Aerosols have both natural origins (*e.g.* dust from soils, wildfires) as well as human origins from combustion (including fossil fuels). Regions that are greening and/or have an increase in rainfall might produce less dust. For anthropogenic aerosols, new ship fuel regulations aimed at reducing sulfur emissions were implemented in three phases, in 2010, 2015 and 2020. Sulfate particles reflect solar radiation. Cleaner air means less solar radiation is reflected, which contributes to surface warming. Indirect effects of sulfate aerosols include increasing the reflectivity of low-level clouds in the subtropics. There is controversy surrounding the importance of this change in sulfate emissions to the Earth's radiation balance (Hodnebrog *et al.*, 2024; Yuan *et al.*, 2024), but the restriction of this effect to major shipping routes suggests a small global impact (Schmidt, 2024).

Changes in clouds are the primary candidate to explain the decline in global albedo since 2015. Two recent papers (Loeb *et al.*, 2024; Goessling *et al.* 2024) have addressed recent variations in cloud properties. By considering satellite and reanalysis data, Loeb *et al.* found that decreases in low- and mid-level clouds since 2015 are the primary reason for decreasing planetary albedo in the Northern hemisphere, whereas in the Southern hemisphere the decrease in planetary albedo is primarily due to decreases in mid-level clouds across all latitude zones. Goessling *et al.* (2024) found that cloud anomalies were mainly due to reduced low-level clouds. Regions with coherent low-level cloud reductions over the past decade include the warm pool region around the Maritime Continent and the northern extratropical western Pacific, as well as large parts of the Atlantic and adjacent land regions. The reduction of global cloud cover identified in these analyses since 2015 is 1-2 percent.

The issue then becomes the cause of the change in cloud cover. Two ~~causes~~ explanations have been posited for the declining cloud cover over the past decade:

- Natural climate variability

- Changes in low cloud cover associated with warming sea surface temperatures, implying an emerging positive feedback to climate change (Hansen and Karecha, 2025)

It is not easy to justify a new positive low cloud feedback that began emerging in 2015. However, there are numerous natural climate signals during this period that are associated with atmospheric circulation changes that can influence the distribution of clouds:

- The 2014-2016 was one of the strongest El Niño events on record
- A cold anomaly beginning in 2015 in the subpolar gyre of the North Atlantic reflects a shift in the ocean circulation pattern associated with decadal variability in the Atlantic (Frajka-Williams *et al.*, 2017; Arthun *et al.* 2021).
- The Pacific Decadal Oscillation positive index peaked in 2016, then declined and has been in negative territory since late 2019
- Eruption of the submarine Hunga-Tonga volcano in 2022

Interannual cloud anomalies associated with the El Niño Southern Oscillation (ENSO) have a significant global signal and strong regional signals, especially over the tropical Indian and Pacific Oceans.

The Hunga Tonga eruption is remarkable in its coincidence with the exceptionally low value of planetary albedo in 2023. Goessling *et al.* (2024) found that the cloud perturbations in 2023 showed a different pattern from the overall pattern since 2015, with reduced cloud cover being most pronounced in the northern hemisphere and tropics. Regional reductions in cloud cover were most pronounced over the eastern Indian Ocean, over South America and extending over the eastern Pacific, as well as over northern North America, in the Southern Ocean around 60S, in the subtropical and eastern North Atlantic, and in parts of the North Pacific.

The Hunga Tonga eruption was unusual in that it injected large amounts of water vapor into the stratosphere in addition to sulfate particles. Early publications focused on the impacts to stratospheric circulations and the direct radiative impact. In 2025, papers are emerging that examine the indirect effects of Hunga Tonga on surface climate using ensemble simulations from earth system models (Bednarz *et al.*, 2025; Kuchar *et al.*, 2025). These papers found statistically significant impacts on regional climates that were driven by circulation changes from couplings between the stratosphere and troposphere. These papers indicate the complex interactions between volcanic activity and climate dynamics. More research is needed to untangle any impacts of the Hunga Tonga eruption on the planetary albedo.

In summary, the decline in planetary albedo and the concurrent decline in cloudiness have emphasized the importance of clouds and their variations to global climate variability and change. A change of 1-2 percent in global cloud cover has a greater radiative impact on the climate than the direct radiative effect of doubling CO₂. While it is difficult to untangle causes of the recent trend, the competing explanations for the cause of the declining cloud cover have substantial implications for assessing the Equilibrium Climate Sensitivity and for the attribution of the recent warming. An additional 10 years of data should help clarify whether this is a strong positive cloud feedback associated with warming, or a temporary fluctuation driven by natural variability

8.5 Attribution of climate impact drivers

The IPCC (Ranasinghe *et al.* 2021) defines “climate impact drivers” or CIDs as “physical climate system conditions (e.g., means, events, extremes) that affect an element of society or ecosystems.” Hence CIDs are those features of the weather and climate system of primary interest in assessing the impacts of climate change since they potentially affect humans and the natural world. For instance, under the heading “Heat and Cold,” CIDs are identified as mean air temperature, extreme heat, cold spells and frost. The IPCC

also points out that CIDs are not necessarily harm-related: depending on the system in question they can be detrimental, neutral, beneficial or a combination.

The first columns Table 12.12 in Ranasinghe *et al.* (2021, p. 1856), reproduced here as Table 8.1, summarize the AR6 assessment of whether a signal of attributable anthropogenic influence has emerged across all major CIDs. One of the themes of this chapter is that attribution methods used by the IPCC tend to overstate the anthropogenic influence and understate the role of natural variability. Nonetheless a striking feature of that summary table is how few CIDs exhibit an anthropogenic signal sufficient to distinguish them from natural variability.

Out of the 33 weather impact categories listed, an anthropogenic signal is asserted with *high confidence* in only five, and with *medium confidence* in a further four. (Note that one of the CIDs is an increase in CO₂ levels, and since it is a tautology to attribute this to increased CO₂ levels this CID can be ignored.) For the rest the IPCC does not claim to have detected anthropogenic drivers. Of the five *high confidence* assertions, two are for changes in average temperatures (air and ocean) hence are not measures of extreme weather. Further, two of the four *medium confidence* assertions are related to ocean chemistry and thus are likewise not related to extreme weather. The IPCC does not assert a human influence on many non-temperature weather features such as wind, precipitation, flooding, or drought.

Other columns of IPCC Table 12.12 report on whether anthropogenic signals are expected to emerge this century under RCP8.5, the most extreme forcing scenario. We have omitted these columns for several reasons. First, because they refer to whether climate models project detectable signals in the observations, which is a very different question than our concern here: whether a signal has been detected in historical data. Second, as we discuss in Chapter 4, the RCP8.5 scenario is a misleading and implausible high-end storyline, it is not a “base case” or business-as-usual projection. Third, the ensemble of detailed models is acknowledged (Palmer and Stevens 2019) to be “not fit for purpose” in describing regional changes and “unable to represent future conditions at the degree of spatial, temporal, and probabilistic precision with which projections are often provided.” (Nissan et al, 2019). Even with these caveats, we note that the omitted columns in AR6 Table 12.12 show most weather impacts are not expected to exhibit an anthropogenic signal through the end of this century.

Climatic Impact-driver Type	Climatic Impact-driver Category	Already Emerged in Historical Period
Heat and Cold	Mean air temperature	1
	Extreme heat	2
	Cold spell	4
	Frost	
Wet and Dry	Mean precipitation	
	River flood	
	Heavy precipitation and pluvial flood	
	Landslide	
	Aridity	
	Hydrological drought	
	Agricultural and ecological drought	
	Fog weather	
Wind	Mean wind speed	
	Severe wind storm	
	Tropical cyclone	
	Sand and dust storm	
Snow and Ice	Snow, glacier and ice sheet	
	Permafrost	
	Lake, river and sea ice	11
	Heavy snowfall and ice storm	
	Hail	
	Snow avalanche	
Coastal	Relative sea level	
	Coastal flood	
	Coastal erosion	
Open Ocean	Mean ocean temperature	
	Marine heatwave	
	Ocean acidity	
	Ocean salinity	13
	Dissolved oxygen	14
Other	Air pollution weather	
	Atmospheric CO ₂ at surface	
	Radiation at surface	

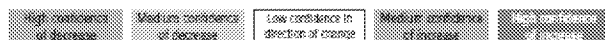


Table 8.1: Reproduction of column 1 of Table 12.12, IPCC AR6 Working Group I report. Emergence of anthropogenic signal in historical period for CIDs shown. White: no signal detected. Blue and orange: change detected (decrease or increase) and confidence level as indicated in color legend.

As discussed in Chapter 6 natural variability dominates patterns of extreme weather systems and simplistic assertions of trend detection are frequently undermined by regional heterogeneity and trend reversals over time. Table 8.1 makes the related point that it is not currently possible to attribute changes in most extreme weather types to human influences. Taking wind as an example, the IPCC claims that an anthropogenic signal has not emerged in average wind speeds, severe windstorms, tropical cyclones or sand and dust storms, nor is one expected to emerge this century even under an extreme emissions scenario. The same applies to drought and fire weather.

8.6 Extreme event attribution (EEA)

Examining the IPCC's treatment of extreme weather and climate event attribution more broadly, AR6 provides an ambiguous assessment of the role of anthropogenic warming that differs between its chapters. Chapter 11 of WG1 states (Seneviratne *et al.*, 2021):

Evidence of observed changes in extremes and their attribution to human influence (including greenhouse gas and aerosol emissions and land-use changes) has strengthened since AR5, in particular for extreme precipitation, droughts, tropical cyclones and compound extremes (including dry/hot events and fire weather). Some recent hot extreme events would have been extremely unlikely to occur without human influence on the climate system.

By contrast, as noted in Section 8.5, Chapter 12 of WG1 (Table 12.12) paints a different picture – presumably, the expert judgment of different groups of authors for the two chapters came to different conclusions (Ranasinghe, 2021):

- High confidence in an increase in extreme heat events in tropical regions where observations allow trend estimation and in most regions in the mid-latitudes, medium confidence elsewhere
- Medium confidence in a decrease in extreme cold events in Australia, Africa and most of northern South America where observations allow trend estimation
- No evidence of emergence in the historical period of a change in river floods, heavy precipitation, drought, fire weather, severe wind storms, and tropical cyclones

While the overall issue of detecting changes in extreme weather events and their attribution remains ambiguous, most of the activity in this area relates to the attribution of particular extreme weather events. The most prominent effort is World Weather Attribution (WWA; worldweatherattribution.org), an international research initiative for extreme event attribution that purports to analyze how climate change influences the likelihood and intensity of extreme weather events. Their approach is to use large ensembles of regional climate models to compare an event in today's climate with one in a counterfactual pre-industrial climate without human influences.

WWA has a prominent public presence in linking extreme weather to climate change, with its press releases attracting considerable attention in public and policy discussions. However, the WWA's extensive promotion of non-peer-reviewed findings, its open admission to shaping analyses to serve litigation, and its methodological challenges have sparked controversies, with critics questioning the robustness and impartiality of their conclusions (Pielke Jr. 2024). Despite these issues, WWA's work continues to influence climate science and media narratives. Technical criticisms of the approach include a lack of a formal detection process; an implicit assumption that 100 percent of the post-industrial warming is caused by greenhouse gases; and a failure to adequately account for internal climate variability.

Because EEA is relatively new, many basic methodological issues have yet to be settled in the expert literature. An important challenge is the lack of data. Extreme events are by definition rare. Many analyses of extreme event types (including the U.S. National Assessment Reports) only evaluate data since 1950 or 1970. However, as emphasized by Chapter 6, many of the worst extreme weather and climate events in U.S. history occurred prior to the first half of the 20th century, including in the early 19th century. And if paleoclimate reconstructions are considered, it becomes very difficult for an event to pass thresholds of what is expected from natural variability, particularly if a reasonably sized geographic region is considered.

Another challenge is defining the event under study. There is a longstanding literature in statistics and econometrics on the challenge of analyzing data with outliers. The issue arises because a data series establishes a probability distribution defining the expected range of observations. If an outlier is observed it might indicate that the underlying process giving rise to the data distribution has changed (which in the weather context would mean that a climate change has been detected) or that the underlying process has multiple regimes each with a different probability distribution, in which case observing an outlier simply means we were temporarily in a different regime, but the system itself was unchanged. If a time series contains only a single outlier event at the end of the series it is not possible to determine which model is the correct one (Chen and Liu 1993). For instance, there might be an “ordinary” weather regime that yields a distribution of summer daytime highs in a particular coastal region, and a second “heatwave” regime that kicks in when an inland blocking event occurs, which yields a temperature distribution centered 15°C higher than the first one. A day with temperatures 13°C above normal would either be an extreme heat anomaly under the first regime or a somewhat cool event under the second, and we have no way in this case of knowing on statistical grounds which view is correct.

Visser and Petersen (2012) and Sardeshmukh *et al.* (2015) both point out that different distributions might fit observed data equally well but yield very different implications about the likelihood of a specific weather event. Visser and Petersen argue that, in view of the deep uncertainties of extreme weather analysis, drawing a connection between individual events and global climate change should be avoided. Furthermore, the existence of an outlier at the end of a data series poses the problem that estimates of the event probabilities will be biased whether the outlier is included or excluded (Barlow *et al.*, 2020). Methods to eliminate the bias have not yet been established, leading some experts (*e.g.* Miralles and Davison 2023) to argue that in settings in which a data series contains a single extreme event at the end, estimation of a return period for the extreme event will be so biased and uncertain that it should be avoided altogether.

We provide a case study of a recent high impact extreme event in the U.S. to illustrate the challenges and ambiguities in attributing the frequency and intensity of extreme weather events to human-caused warming.

8.6.1 Case study – 2021 Western North America heat wave

The 2021 Western North America heat wave was an extreme event that affected much of Western North America in late June 2021. Surface temperature records were set in Portland, OR (116°F; previous record 107°F) and Seattle, WA (108°F; previous record 103°F) (Mass *et al.*, 2024).

The WWA team generated international headlines with their analysis, which provided the following attribution statements (WWA, 2021; Philip *et al.*, 2022):

- Based on observations and modeling, the occurrence of this heatwave was virtually impossible without human-caused climate change.
- The event is estimated to be about a one in 1000-year event in today’s climate.
- The event would have been at least 150 times rarer without human-induced climate change.
- This heatwave was about 2°C hotter than it would have been if it had occurred at the beginning of the industrial revolution (when global mean temperatures were 1.2°C cooler than today).

But an important counter to the first claim is that other researchers concluded from historical weather data that while a heat wave of the magnitude observed was indeed virtually impossible without anthropogenic climate change, it was also virtually impossible *with* climate change. Bercos-Hickey (2022) noted “these temperatures were virtually impossible under any previously experienced meteorological conditions, with or without global warming.” McKinnon and Simpson (2022) stated “the most likely explanation remains that the weather event itself was ‘bad luck.’” The 2023 Oregon Climate Assessment (Fleischman, 2023) concluded that the heat dome would have formed even without climate change and “There is no evidence that the highly unusual combination of weather features that drove the heat dome were made more likely by climate change, and climate models do not project an increase in the frequency of high-pressure ridges over the Pacific Northwest” (Fleischman, 2023, p. 49).

Mass *et al.* (2024) summarizes the proximate sequence of compound events leading to the heat wave. There was a record-breaking mid-tropospheric ridge over the Pacific Northwest, forced by a tropical disturbance in the western Pacific. This produced record-breaking mid-tropospheric temperatures, strong subsidence in the lower atmosphere, low-level easterly flow that produced downslope warming on regional terrain and the removal of cooler marine air, and an approaching low-level trough that enhanced downslope flow. The event occurred at a time of maximum solar insolation, and drier-than-normal soil moisture. Using a storyline approach, Mass *et al.* assessed that there was no trend in drought and dry soils in the Pacific Northwest; there is no evidence of global warming producing stronger ridges of high pressure, and no observed trend in heat waves or record temperatures in the region. They concluded that although anthropogenic warming might have contributed as much as 2°F to the magnitude of the event, there is little evidence of further amplification of the event from increasing greenhouse gases.

Bercos-Hickey *et al.* (2022) conducted a statistical analysis of a model-based attribution study of this heat wave. Because the event is a far outlier and far above the bounds of Generalized Extreme Value distributions fitted from historical data, they concluded that estimates of return times, quantitative changes in event magnitude and frequency, and probability of the extreme temperatures such as provided by the WWA are not accurate and should be interpreted as having low confidence. They found that hindcast attribution methods using an ensemble of regional climate models, combined with Pearl (2009) causal inferences, can provide limited and conditional information about the magnitude of the human influence on this heatwave - they provided a more highly constrained estimate that human activities caused a ~1.4°F–1.8°F increase in the daily maximum temperatures.

Zeder *et al.* (2023) also concluded that the methods employed by Philip *et al.* (2022, the WWA analysis) tend to overstate the rarity of extreme heat waves, leading to a biased perception of the effect of climate change on the heatwave event: “The tendency to overestimate the return period of observed extreme heatwave events may fuel the impression that seemingly impossible heatwave extremes are currently clustering at an unprecedented rate.”

The 2021 Western North America heatwave was a rare and unprecedented compound weather event that broke century-old temperature records by as much as 9°F. While the WWA team received worldwide publicity for their rapid attribution claims blaming anthropogenic climate change, subsequent peer-reviewed analyses showed that the event was caused by rare meteorological conditions that were not made more probable by global warming.

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PART III: IMPACTS ON ECOSYSTEMS AND SOCIETY

9 CLIMATE CHANGE AND U.S. AGRICULTURE

Chapter summary

There has been abundant evidence going back decades that rising CO₂ levels benefit plants, including agricultural crops, and that CO₂-induced warming will be a net benefit to U.S. agriculture. The warming since 1981, in combination with adaptive behavior by farmers, has been beneficial for maize. The increased ambient CO₂ has also boosted productivity of all major U.S. crop types. Various attempts to argue that CO₂-induced warming will harm agriculture do not hold up under scrutiny.

9.1 Econometric analyses

Econometric analyses of the effects of climate change on agriculture have sought to integrate information about long term crop yield changes in response to temperature and precipitation changes under assumptions about adaptive behavior by farmers. One method focuses on variations in agricultural land values. The rationale is that if climate change is a long-term net benefit for agriculture it should be capitalized into higher market values for agricultural land, and vice versa. While individual crops might benefit from, or be harmed by, climate change, once the adaptive responses of farmers are considered the value of farmland represents an index of whether the changes are expected to be beneficial or not. Mendelsohn *et al.* (1994) examined the relationship between historical climatic variations on agricultural land values and concluded global warming would be slightly beneficial to American agriculture.

The Mendelsohn *et al.* method, called Ricardian analysis after David Ricardo, the 19th century British economist who pioneered the study of land values, attracted subsequent criticism from authors who argued it failed to account for differences in land values attributable to fixed locational characteristics like soil quality, and non-climatic changes including nearby urbanization. Deschênes and Greenstone (2007) looked at agricultural profits instead of land values and reached conclusions similar to Mendelsohn *et al.* (1994), namely that past climatic variations had relatively little effect on farm profitability and that warming would likely yield small overall benefits for the U.S. agricultural sector. However, a subsequent exchange with critics led them (Deschênes and Greenstone 2012) to revise their conclusions and project potentially large losses in U.S. agriculture due to climate warming.

Burke and Emerick (2016) looked at temperature variations over 1980 to 2000 and argued that farmers were not as able to adapt to temperature changes as the Ricardian method assumes and that climate change would have large negative impacts on corn and soy yields. Schlenker and Roberts (2009) similarly argued that yield gains to past warming would not carry over to the future and corn and soy yields would sharply decrease this century due to climate change.

Two recent studies have argued that pessimistic findings such as these are not robust. Ortiz-Bobea (2019) argued that land values aggregate farm and non-farm influences and the latter need to be filtered out. He developed a data set using cash rents for agricultural activity as a measure of land value specifically for farming activity. Whereas the land value model implied future losses under climate warming, the same model estimated using cash rents did not, leading the author to conclude the pessimistic results were due to using an inaccurate measure of the returns to farming activity. Bareille and Chakir (2023) assembled a large data base on farm sale prices in France for properties that sold twice between 1996 and 2019. They could replicate pessimistic results showing negative effects of warming on agricultural land values using conventional econometric modeling. But by taking advantage of the repeat sales data, which provides information on site-specific changes in land prices, they found the results reversed and implied that climate change will be very beneficial for French agriculture. The authors concluded that, taking adaptation into account, a warming climate would yield positive benefits for French agriculture that were between two and

20 times larger than had previously been estimated. On average, with full adaptation, they concluded that climate changes under the medium RCP4.5 scenario could double the value of French farmland by 2100.

A major deficiency of all these studies, however, is that they omit the role of CO₂ fertilization. Climate change as it relates to this report is caused by GHG emissions, chiefly CO₂. The econometric analyses referenced above focus only on temperature and precipitation changes and do not take account of the beneficial growth effect of the additional CO₂ that drives them. As explained in Chapter 2, CO₂ is a major driver of plant growth, so this omission biases the analysis towards underestimation of the benefits of climate change to agriculture.

9.2 Field studies of CO₂ enrichment

One of the ways the effect of CO₂ on crop growth has been studied is through “free air enrichment experiments” or FACE plots, in which small sources of CO₂ are placed in fields surrounding plants and the growth response to elevated CO₂ under varying weather conditions are recorded. Ainsworth *et al.* (2020) summarizes results from about 250 such studies. They found that elevation of CO₂ by 200 ppm caused an average 18 percent increase in crop yield in C3 plants. C4 plants exhibited benefits mainly under drought conditions.

In addition to FACE plot experiments there have been thousands of laboratory experiments on the effects of CO₂ on all kinds of plant growth. Here we review some of the results on key U.S. agricultural crops.

Soybean

Studies on the impact of elevated CO₂ on Soybean (*Glycine max*(L.) Merr.) plants in the water-deficient region of Huang-Huai-Hai Plain, China, showed that elevated CO₂ concentrations improved photosynthesis rate, water use efficiency, and growth Li (2013) under both normal conditions and drought conditions. The CO2Science.org website reports on 108 published experiments between 1985 and 2019 exposing soybean to enriched CO₂ levels. Converted to a +300ppm common scale the average growth benefit was +50.9 percent. There were also ten studies reporting on +600 ppm CO₂ enrichment, which increased photosynthesis by an average of 90.3%.

Maize (corn)

The CO2Science.org website reports on 28 published experiments between 1983 and 2018 exposing corn (*Zea mays* L.) to enriched CO₂ levels. Converted to a +300ppm common scale the average growth benefit was +23.7 percent. Corn also benefits from increased drought tolerance under elevated CO₂. An experimental study (Allen Jr., 2011) exposed plants to water stress conditions in sunlit controlled-environment chambers at 360 ppm (ambient) and 720 ppm (elevated) CO₂. The drought stress caused a 41 percent loss of growth under ambient CO₂ but only a 13 percent loss under elevated CO₂.

Wheat

The CO2Science.org website reports on 92 published experiments between 1983 and 2020 exposing common wheat (*Triticum aestivum* L.) to enriched CO₂ levels. Converted to a +300ppm common scale the average growth benefit was +67.6 percent. Blandino (2020) measured both the yield and nutritional quality of an “improver” hybrid wheat variety and its parents under elevated CO₂ levels (+166 ppm). They reported a grain yield increase of +16 percent but a 7 percent decrease in grain protein levels. But they also found that the food quality of different wheat varieties responded differently to elevated CO₂ levels, so that with proper varietal selection, growers could select wheat types that best take advantage of the elevated CO₂.

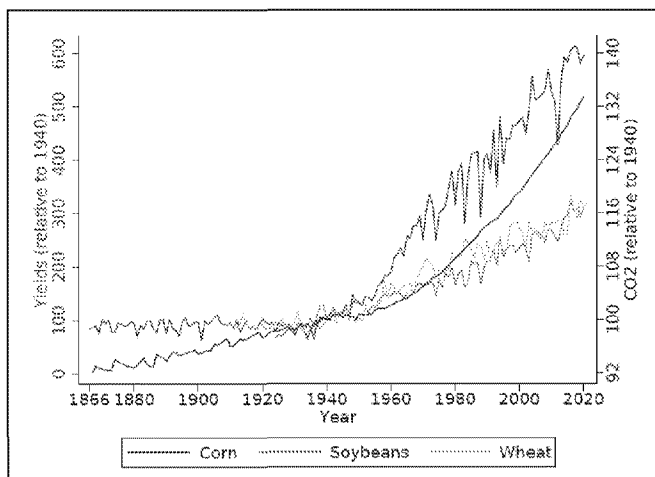


Figure 9.1: U.S. average CO₂ levels and yields of corn, soy and wheat all normalized so 1940=100. Source: Taylor and Schlenker (2021)

Further evidence

A 2021 report from the U.S. National Bureau of Economic Research (Taylor and Schlenker 2021) used satellite-measured observations of outdoor CO₂ levels across the United States, matched to county-level agricultural output data and other economic variables. After controlling for the effects of weather, pollution and technology the authors concluded that CO₂ emissions had boosted U.S. crop production since 1940 by 50 to 80 percent, attributing much larger gains than had previously been estimated using FACE experiments. They found that every ppm of increase in CO₂ concentration boosts corn yields by 0.5 percent, soybeans by 0.6 percent, and wheat by 0.8 percent.

Beyond growth benefits, extra CO₂ boosts plant resilience to dryness. Deryng *et al.* (2016) surveyed evidence on crop water productivity (CWP, the yield per unit of water used), drawing attention to the potential for CO₂ both to enhance photosynthesis and reduce leaf-level transpiration (water loss during leaf respiration). They surveyed all available FACE data on crop yield changes for maize (corn), wheat, rice and soybean, and combined it with crop model data simulating yield responses as of 2080 under the extreme RCP8.5 scenario in four growing regions (Tropics, Arid, Temperate and Cold) each of which were split into rainfed and irrigated sub-regions. In every region they reported that models without CO₂ fertilization predicted CWP losses but CO₂ fertilization more than offset them; once fertilization was taken into account, all regions showed a net CWP gain. Deryng *et al.* also reported that negative yield impacts on wheat and soybean were fully offset by CWP gains and mitigated by up to 90 percent for rice and 60 percent for maize. Similarly, Cheng *et al.* (2017) noted that increased global plant growth from 1982 to 2011 due to rising CO₂ uptake was accompanied by such large gains in CWP that despite the extra biomass global water use by plants had not increased.

Deryng *et al.* (2016) assumed that climate change would “exacerbate water scarcity”. It is noteworthy that while models predict drylands will expand under climate warming, observations show the opposite: greening is happening even in arid areas. Zhang *et al.* (2024) report that due to increased CO₂ levels “increasing aridity in drylands won’t lead to a general loss of vegetation productivity”; increased desertification will happen in at most 4 percent of currently arid areas.

9.3 Crop modeling meta-analyses

Notwithstanding the abundant evidence for the direct benefits of CO₂ and of CO₂-induced warming on crop growth, in 2023 the U.S. Environmental Protection Agency (EPA 2023) boosted its estimate of the Social Cost of Carbon (SCC) about five-fold based largely on a very pessimistic 2017 estimate of global agricultural damages from climate warming (Moore *et al.*, 2017). One of the two damage models used by the EPA attributed nearly half of the 2030 SCC to projected global agricultural damages based on the Moore *et al.* (2017) analysis. This study was a meta-analysis of crop model studies simulating yield changes for agricultural crops under various climate warming scenarios. Moore *et al.* projected declining global crop yields for all crop types in all regions due to warming.

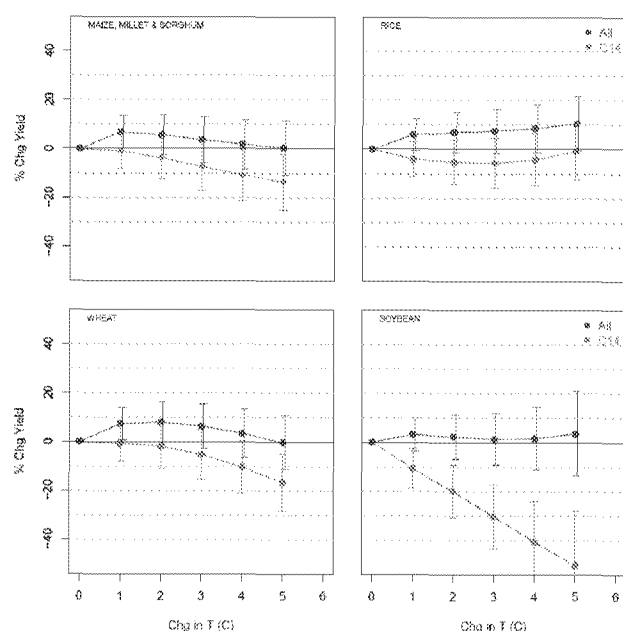


Figure 9.2 Crop yields under CO₂-induced climate warming. Blue: as published in Moore *et al.* (2017). Green: after including omitted data. Source: McKittrick (2025).

McKittrick (2025) re-examined the Moore *et al.* database and found that, while it claimed to cover 1,722 studies, only half the entries ($N=862$) had complete records, so that the sample available for regression analysis was much smaller than both studies indicated. McKittrick noted that the records most commonly missing were the changes in ambient CO₂ and found that in many cases these could be recovered from the underlying studies or the original climate scenario tables, thereby increasing the usable sample size by 40 percent. The crop yield projections incorporating the newly available data changed considerably. As shown in Figure 9.2, whereas the partial data set implied warming would decrease yield (blue lines), the complete data set implied constant or increase global yields, even out to 5°C warming (green lines).

9.4 CO₂ fertilization and nutrient loss

Evidence has shown that CO₂-induced biomass gains are sometimes accompanied by reductions in the concentrations within the grains of protein and other key nutrients such as iron and zinc (Ebi *et al.* 2021). Some experimental evidence has shown that the rising temperatures expected to accompany higher CO₂ levels will offset this loss (Köhler *et al.* 2019) although the evidence for this is mixed, as is the evidence that nutrient dilution observed to date is attributable to higher CO₂ (Ziska 2022). If nutrient dilution does occur under rising CO₂ levels there are several adaptive strategies that can be pursued.

First, selective breeding for resilience to dilution and higher micronutrient content is already long established (Saltzman *et al.* 2017) and has proven to be a cost-effective agronomic strategy (Ebi *et al.* 2021). Strategies can include both conventional breeding and use of genetically-modified organisms. An example of the latter is Golden Rice, which contains elevated levels of beta-carotene to boost biosynthesis of vitamin A in the human body. Optimal strategies will be location-specific because they vary by crop, climate and soil type (Ebi *et al.* 2021). Second, fortification of food products with micronutrients is already routine. Folic acid (a B vitamin) is added to flour and many other foods; iodine is added to table salt; most commercial breakfast cereals are fortified with iron and numerous vitamins, *etc.* Third, dietary supplements in the form of multivitamin tablets are inexpensive, widely-available and routinely consumed.

One concern about reliance on adaptive strategies is whether they are feasible in low-income countries. Micronutrient deficiency is already a problem in the developing world and dietary supplements have proven to be an effective low-cost response (Ebi *et al.* 2021). It should also be noted the IPCC emission scenarios that generate high levels of warming do so on the basis of high levels of income growth. The SSP scenarios³ assume that, compared to 2005 levels, global per capita income will double by 2100 in the lowest growth case (SSP3), and in the highest emission case (SSP5) global per-capita income will grow nearly 16-fold. In that scenario even the poorest regions (Africa and the Middle East) end up with a per capita income of about US\$126,000, 70 percent higher than 2020 U.S. per capita income (about US\$75,000). Consequently the same scenarios in which CO₂ levels increase the most are also those in which global poverty is largely eliminated, in which case all countries will be able to afford dietary supplements as necessary to address micronutrient deficiencies, if they arise and cannot be addressed using on-farm agricultural strategies.

In summary, there is abundant evidence going back decades that rising CO₂ levels benefit plants, including agricultural crops, and that CO₂-induced warming will be a net benefit to U.S. agriculture. To the extent nutrient dilution occurs there are mitigating strategies available that will need to be researched and adapted to local conditions.

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³ See compilation at [HYPERLINK "https://ourworldindata.org/explorers/ipcc-scenarios?Metric=GDP&Sub-metric=Total+radiative+forcing&Region=Global&country=SSP1+-+Baseline-SSP2+-+Baseline-SSP3+-+Baseline-SSP4+-+Baseline-SSP5+-+Baseline"]

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10 MANAGING RISKS OF EXTREME WEATHER

Chapter summary

Trends in losses from extreme weather and climate events are dominated by population increases and economic growth. Technological advances such as improved weather forecasting and early warning systems have substantially reduced losses from extreme weather events. Better building codes, flood defenses, and disaster response mechanisms have lowered economic losses relative to GDP. The U.S. economy's expansion has diluted the relative impact of disaster costs, as seen in the comparison of historical and modern GDP percentages. Heat-related mortality risk has dropped substantially due to adaptive measures including the adoption of air conditioning, which relies on the availability of affordable energy. U.S. mortality risks even under extreme warming scenarios are not projected to increase if people are able to undertake adaptive responses.

10.1 Socioeconomic context

Risks from human-caused climate change are affected by natural weather and climate variability and are dominated by the exposure of wealth in coastal and other disaster-prone regions and vulnerabilities of poorer populations. The evolution of climate risk in the U.S. has been dominated by societal factors, rather than by changes to the actual weather and climate hazards. Deaths from weather disasters have decreased substantially since 1900, even as U.S. population grew from 76 million in 1900 to over 331 million in 2020 (Goklany 2011). For example, the Galveston hurricane killed over 8,000 people in 1900 (0.01 percent of the U.S. population), whereas the worst recent disaster, Hurricane Katrina in 2005, killed 1,800 people (or 0.0006 percent of the U.S. population) (NOAA National Hurricane Center 2025; U.S. Census Bureau 2025).

Technological advances have substantially reduced losses from extreme weather events. Early warning systems, satellite monitoring, and improved weather forecasting have reduced deaths, although exact numbers are hard to quantify (Deryugina and Hsiang 2023). U.S. weather forecasting has been estimated to reduce losses from weather events with an annualized benefit of \$31.5B, protecting lives, property, and supporting agriculture and transportation (NRC 2010). Improved hurricane forecasts have reduced pre- and post-landfall spending, with annual per hurricane cost reductions estimated at \$5B (Molina and Rudik 2024).

Infrastructure improvements have contributed to substantial reductions in losses from extreme weather events. Building codes, such as those implemented in Florida after Hurricane Andrew (1992), have reduced losses by ensuring structures can withstand high winds and floods. Homes built after 2002 showed minimal damage during Hurricane Michael (2018), unlike older homes (FEMA 2020). Sea walls, like the Galveston Seawall, protect against wave action and storm surge. The New Orleans Hurricane and Storm Damage Risk Reduction System successfully mitigated storm surge during Hurricane Isaac (2012) (Battelle Memorial Institute 2013). Inland dams in the U.S. help control flooding by storing excess water during

heavy rains. It is estimated that the Tennessee Valley Authority (TVA) dams prevent about \$309M in annual flood damage in the TVA region and along the Ohio and Mississippi Rivers (TVA, 2025). During Hurricane Helene (2024), TVA’s strategies prevented approximately \$406M in potential damages (APPA 2024).

10.2 Data challenges

Since 1980, NOAA has provided a count of annual U.S. weather-related disasters that it estimates to have exceeded \$1 Billion (inflation adjusted), showing a substantial increase starting in 2008. NOAA and other government officials have cited the upward trend in the Billion Dollar Disaster series as evidence that climate change is making extreme weather worse (Pielke Jr., 2024). But over time, population and wealth have increased dramatically in the U.S., so when an extreme weather or climate event occurs, there is more damage even if there is no underlying trend in the frequency or intensity of extreme weather. Pielke Jr. (2024) demonstrates that losses per weather disaster as a proportion of GDP have decreased by about 80 percent since 1980, as shown in Figure 10.1. Pielke Jr. (2024) also argued that in addition to relying on opaque data sources and unreported adjustments, NOAA failed to normalize its Billion Dollar Disaster data series properly for changes in population exposure and wealth. In May 2025 NOAA announced it has withdrawn the Billion Dollar Disaster product from publication (Pielke Jr., 2025).

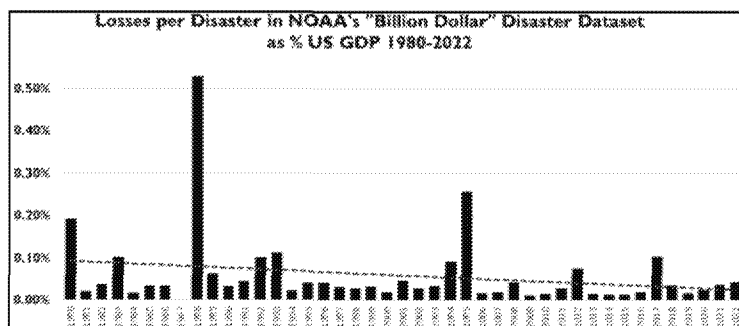


Figure 10.1: Losses per disaster as a % of Gross Domestic Product in NOAA's billion-dollar disaster dataset (the version downloaded in July 2023), 1980 to 2022. Source: (Pielke, Jr. 2024)

In summary, trends in losses from extreme weather and climate events are dominated by population increases and economic growth. Technological advances such as improved weather forecasting and early warning systems have substantially reduced losses from extreme weather events. Better building codes, flood defenses, and disaster response mechanisms have lowered economic losses relative to GDP. Finally, the U.S. economy's expansion has diluted the relative impact of disaster costs, as seen in the comparison of historical and modern GDP percentages.

10.3 Mortality from temperature extremes

10.3.1 Heat and cold risks

Changes in temperature extremes are among the most certain impacts expected in a warming world. It stands to reason that extreme heat events would likely become more frequent while extreme cold events would become less frequent. This pattern is evident in the historical period, though not in the continental U.S. (Chapter 6), and is expected to continue with further warming.

Mortality during heat extremes is typically caused by heat stroke and heat exhaustion, while mortality during cold extremes typically stems from hypothermia and heart strain. Global mortality is substantially greater for cold conditions than for hot conditions (Zhao *et al.* 2021, Ritchie 2024). Unlike with heat-related mortality, cold-weather risks set in even at moderately cold conditions (Gasparini *et al.* 2015, Lee and Dessler 2023). The U.S. EPA (citing data from the Centers for Disease Control) reports that on average over 1999 to 2015 there were 2.2 deaths per million Americans for which cold was listed as the main underlying cause and an additional 2.4 deaths per million for which cold was listed as a contributing factor (EPA 2025). By contrast there were 1.3 deaths per million for which heat was the main cause and an additional 0.8 per million for which it was a contributor. By this metric cold accounts for approximately double the weather-related deaths as does heat.

Epidemiological methods that consider correlational evidence, not just death certificate reports, indicate the cold/warm ratio might be much higher. A 13-country study of 74 million deaths from 1985 to 2012 estimated that, on average, 7.7 percent of deaths were attributable to sub-optimal temperatures, of which 7.4 percent were attributable to cold and only 0.4 percent were attributable to heat (Gasparini *et al.* 2015). In other words, cold weather killed 18.5 times as many people as did hot weather.

Figure 10.2 shows the distribution of results from Gasparini *et al.* (2015) by country. For the United States the fraction of deaths attributable to temperature was 5.9 percent, of which 5.5 percent was due to cold, thus cold weather killed 14 times as many people as hot weather.

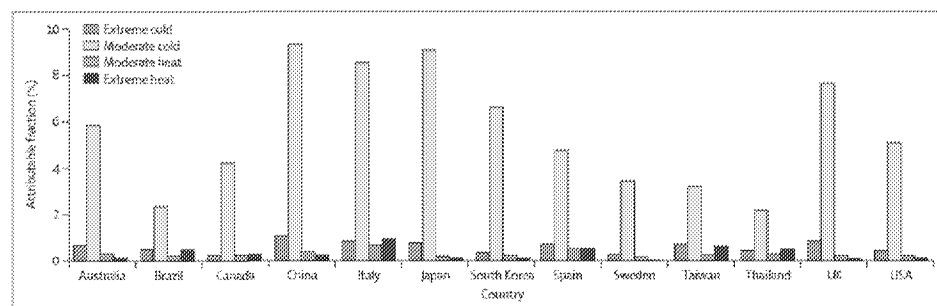


Figure 10.2. Mortality attributable to extreme and moderate cold and heat by country. Source: reproduced from Gasparini *et al.* (2015).

There is strong evidence that people adapt to weather risks. Lee and Dessler (2023) reported that 86 percent of temperature-related deaths across 40 cities in the U.S. were due to cold-related mortality, and that due to adaptation the relative risk of death declined in hot and cold cities alike as seasonal temperatures increased. Allen and Sheridan (2018) found that short, early-season cold events were 2 to 5 times deadlier

than hot events, but the mortality risk of both cold and hot extremes drops to nearly zero if the events occur late in the season.

Davis *et al.* (2003) examined heat-related mortality in 28 U.S. cities from the 1960s to the end of the 1990s and found that heat-related mortality declined by three-quarters over the sample period. Bobb *et al.* (2014) examined mortality data for 106 million people in over 100 cities U.S. cities and found a 60 percent decline in average heat-related mortality over the period 1987–2005, from 51 per thousand deaths to 19. They furthermore found that the greatest drop was among seniors over the age of 75. In a study of 42 million deaths in 211 U.S. cities from 1962 to 2006, Nordio *et al.* (2015) found a more than 90 percent decline in the risk of mortality from excess heat.

In the context of large declines in heat-related mortality, rising temperatures are associated with a net saving of lives since they reduce mortality from cold events. The IPCC AR6 Working Group 2 Chapter 16.2.3.5 (O'Neill *et al.* 2022) acknowledges that heat-related mortality risk is declining over time:

Heat-attributable mortality fractions have declined over time in most countries owing to general improvements in health care systems, increasing prevalence of residential air conditioning, and behavioral changes. These factors, which determine the susceptibility of the population to heat, have predominated over the influence of temperature change.

Yet the IPCC misrepresents the overall situation in its AR6 Synthesis report. Section A.2.5 of that document states: “In all regions increases in extreme heat events have resulted in human mortality and morbidity (*very high confidence*).” But it is silent on the larger decline of deaths during extreme cold events.

The observed decline in U.S. heat-related mortality has been specifically attributed to adaptation. Wang *et al.* (2018) exploited the spatial variability of heatwave-related mortality across 209 U.S. cities from 1962 to 2006. While simple correlation appeared to imply an increase in mortality risk during heatwaves, accounting for adaptation to heatwave intensity caused the effect to fall to near zero and become statistically insignificant. They used the results of their epidemiological model to project heat-related mortality out to 2050 under four RCP warming scenarios (including RCP8.5) with and without adaptive behavior. Assuming people continue to adapt to the heatwave risks in the regions in which they live, Wang *et al.* (2018) project not only no increase in heat-related mortality, but an overall mortality *decrease* for the U.S. as a whole. They conclude that

Ignoring adaptation would result in a substantial overestimate of future mortality related to heat waves... Accounting for adaptation, the overall heat-related mortality by 2050 would not change substantially over time compared to 2006.

10.3.2 Mortality risks and energy costs

A 2016 study of U.S. long term mortality risks associated with temperature variations (Barreca *et al.* 2016) showed that increases in mortality across the U.S. are associated with both cold weather and hot weather. But over time, the introduction of electricity and the adoption of central heating and air conditioning (AC) dramatically reduced both risks, especially those associated with hot weather. Prior to 1960, a day above 90°F (32°C) added 2.2 percent to the average mortality risk rate, but after 1960 the same weather added only 0.3 percent to mortality risk, an 85 percent reduction. Prior to 1960 temperatures below 39°F (4°C) added about 1 percent to mortality risk but after 1960 the same weather only added about half

that amount. Adaptation through conventional household improvements dramatically reduced public vulnerability to weather extremes. The entire reduction in hot weather mortality was attributable to widespread adoption of indoor AC, which depended on the availability of reliable and affordable electricity.

The corollary of this finding is that the use of heating and cooling systems depends on energy being affordable. Doremus *et al.* (2022) showed that wealthy and poor households in the U.S. adjust their energy expenditures at similar rates in response to moderate temperature swings, but not in response to extreme temperature swings. When temperatures swing to very cold levels (< 5°C) energy spending in high-income households rises by 1.2 percent but in low-income households by only 0.5 percent. On very hot days (>30°C) electricity spending in high-income households rises by 0.5 percent but does not change at all in low-income households. The latter result is observed even in subsamples where all households have AC. The implication is that even with widespread adoption of home heating and cooling systems, inability to afford energy leaves low-income households exposed to weather extremes.

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11 CLIMATE CHANGE, THE ECONOMY, AND THE SOCIAL COST OF CARBON

Chapter summary

Economists have long considered climate a relatively unimportant factor in economic growth, a view echoed by the IPCC itself in AR5. Mainstream climate economics has recognized that CO₂-induced warming might have some negative economic effects but they are too small to justify aggressive abatement policy and that trying to “stop” or cap global warming even at levels well above the Paris target would be worse than doing nothing. An influential study in 2012 suggested that global warming would harm growth in poor countries, but the finding has subsequently been found not to be robust. Studies that take full account of modeling uncertainties either find no evidence of a negative effect on global growth from CO₂ emissions or find poor countries as likely to benefit as rich countries.

Social Cost of Carbon (SCC) estimates are highly uncertain due to unknowns in future economic growth, socioeconomic pathways, discount rates, climate damages, and system responses. The SCC is not intrinsically informative as to the economic or societal impacts of climate change. It provides an index connecting large networks of assumptions about the climate and the economy to a dollar value. Some assumptions yield a high SCC and others yield a low or negative SCC (i.e. a social benefit of emissions). The evidence for or against the underlying assumptions needs to be established independently; the resulting SCC adds no additional information about the validity of those assumptions. Consideration of potential tipping points does not justify major revisions to SCC estimates.

11.1 Climate change and economic growth

11.1.1 Overview

It has long been noted that economies tend to do poorly in very cold and very hot regions, with the optimum somewhere in between (Nordhaus, 2006). Temperature-sensitive economic activity migrates, whenever possible, to where it is best suited, and society adapts to the local climate. Based in part on these observations, in 1992 Thomas Schelling, then President of the American Economic Association, argued that any effects of climate change on U.S. economic activity would be small relative to the many other changes that would happen (Schelling 1992).

... Manufacturing rarely depends on climate, and where temperature and humidity used to make a difference, air conditioning has intervened. When Toyota chooses among Ohio, Alabama, and Southern California for locating an automobile assembly, geographical considerations are important, but not because of climate... Finance is little affected by climate; similarly for health care, or education, or broadcasting. Transportation can be affected, but improvements in all-weather landing and take-off in the last 30 years are greater than any differences that climate makes. If the average effect is a warming, iced waterways and snow removal may decline in importance. Construction is affected, mainly by cold, and if the average effect is in the direction of warming construction may benefit slightly.

It is really agriculture that is affected. But even if agricultural productivity declined by a third over the next half century, the per capita GNP we might have achieved by 2050 we would achieve only in 2051. ...

I conclude that in the United States, and probably Japan, Western Europe, and other developed countries, the impact on economic output will be negligible and unlikely to be noticed.

Thirty years later virtually the identical point was made by the IPCC itself in the Fifth Assessment Report (Arent *et al.* 2014, emphasis added)

For most economic sectors, the impact of climate change will be small relative to the impacts of other drivers. Changes in population, age, income, technology, relative prices, lifestyle, regulation, governance, and many other aspects of socioeconomic development will have an impact on the supply and demand of economic goods and services that is large relative to the impact of climate change.

Evidence since AR5 does not change this assessment. Mohaddes *et al.* (2023) found that warm weather shocks have small negative effects on U.S. state-level output but not income, while cold weather shocks negatively affect both and the impacts are about four times larger, implying a shift to warmer conditions would, if anything, yield a net economic benefit for the U.S.

These statements are validated by experience. Since 1900, the average global surface temperature anomaly warmed 1.3°C, about as much as the IPCC predicts will occur in the next century under a moderate emissions scenario. But even as the globe warmed and the population quintupled, humanity prospered as never before. For example, global average lifespan went from thirty-two years to seventy-two years, economic activity per capita grew by a factor of seven, and the death rate from extreme weather events plummeted by a factor of fifty.

While extreme weather events are costly, in all modern economies they are becoming steadily less and less important (Formetta and Feyen, 2019). Since 1990 weather-related disaster losses have declined as a

proportion of global GDP (Pielke Jr , 2018, 2020) as have mortality risks (Formetta and Feyen, 2019). While economic weather-related insurance payouts are rising, this is fully explained by the growth in the size of the economy and the value of the insured asset stock. Past increases in episodes extreme weather have not had a significant effect (positive or negative) on the market value of insurance firms (Hu and McKittrick, 2015). Nor have past extreme weather events had a significant effect on U.S. banks' performance (Blickle *et al.*, 2021); warming has even been shown to be beneficial for the finance and insurance sector (Mohaddes *et al.* 2023). Figure 11.1 below is illustrative. For these reasons, economists have long been reluctant to endorse attempts to “stop” climate change or even aggressively reduce GHG emissions because the costs would not be worth it. As one critic of the economics of climate policy put it (Storm 2017):

Mainstream climate economics takes global warming seriously, but perplexingly concludes that the optimal economic policy is to almost do nothing about it... The contrast is striking. While climate science is sending out loud-and-clear messages that fossil-fuel disinvestment must start now, letting go of coal and oil and diverting resources into renewable energy technology systems, to keep warming below the 2°C limit (IPCC 2014), mainstream climate economics claims that overly ambitious climate targets will unnecessarily hurt the economy and immediate de-carbonization is too expensive. Most climate economists thus recommend humanity to just wait-and-see.

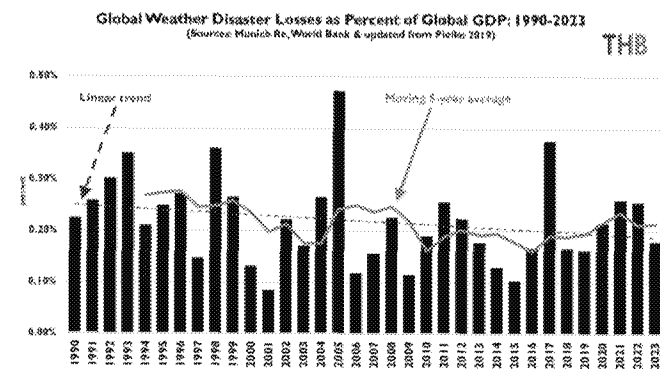


Figure 11.1 Global weather losses as a fraction of GDP. Source: Pielke Jr. (2023)

The mainstream economics position on the climate policy question is best represented by the findings over the past three decades from Integrated Assessment Models (IAMs) of climate change policy, for which Yale economist William Nordhaus was awarded the 2018 Nobel Memorial Prize in economics. IAMs combine economic, climate, and social data into a unified framework for simulating climate damages and evaluating the optimal policy response (Resources for the Future, 2025). Nordhaus' work has generally supported modest global climate policy, with aggressive measures largely deferred to later in the century. Nordhaus developed the so-called Dynamic Integrated Model of the Climate and Economy model, or DICE, in the early 1990s (Nordhaus, 1993) to study the interaction between climate change, climate policy, and global economic growth over long times. DICE assumes global emission control can be coordinated at no

cost and asks what the optimal policy target should be. The climate component of the DICE model is based on simplified climatological modeling. The version of DICE at the time of Nordhaus' Nobel Prize award was described in Nordhaus (2018); it had a climate sensitivity parameter of 3.1°C for CO₂ doubling, consistent with the best estimate in IPCC reports (3.0°C).

The baseline (no policy) scenario incurs only \$0.4T in global abatement costs and leads to \$134.2T (trillion dollars) in global climate damages for a total cost of \$134.6T. The climate model component of DICE projects 4.1°C warming by 2100 relative to preindustrial temperatures. Note that this is a higher estimate of warming than many IPCC climate models.

The Optimal Policy scenario barely deviates from the business-as-usual path. It aims for +3.5°C warming, in other words we modestly scale back fossil fuel use and otherwise just live with almost all the warming. This suggests that most CO₂ emissions are less harmful than the policies that would be necessary to abate them. Trying to prevent warming causes costs quickly to outstrip the benefits. Pursuing a goal of capping warming at 2.5°C creates total costs of \$177.8T, which is \$43.2T worse than doing nothing at all. Nordhaus didn't evaluate trying to reach a Paris-type target of 1.5°C warming but it would be even more costly.

A subsequent edition of DICE includes higher assumed damages from warming which, not surprisingly leads to more aggressive policy recommendations, as explained in the section below on the Social Cost of Carbon.

11.1.2 Empirical analysis of climate change and economic growth

Many other studies have used econometric methods applied to historical data, instead of IAMs, to study the potential impact of climate change on economic growth. Dell *et al.* (2012, herein DJO12) was an influential study that used a multi-country panel of national-level data spanning 1950 to 2005, in which they matched climate and economic data by averaging temperature from the local grid cell level up to the national level using population weights. They found that warming yields an insignificant positive effect on income growth in rich countries but a significant negative effect in poor countries. Moore and Diaz (2015) modified the DICE model to take that finding into account and concluded that it implied a dramatically higher Social Cost of Carbon due to the compounding effects of income loss over time.

A large subsequent literature has debated the robustness of the DJO12 findings. Burke *et al.* (2015) analyzed a global panel with temperatures averaged up to the national level and found a negative effect on growth from warming in rich and poor countries alike when the national average temperature is above 13°C. Zhao *et al.* (2018) used the G-Econ data set of Nordhaus (2006), which breaks economic activity down to the grid cell level, replicated DJO12-type results on the same subset of countries as used in DJO12 but then showed that on the full global sample warming increases growth in rich countries and poor countries alike, though the positive effect in the latter group is confined to where local temperatures are below about 16°C. Greßer *et al.* (2021) developed a regional economic data set for 1,542 sub-national regions around the world between 2005 and 2015 and found temperature had no effect on growth. Yang *et al.* (2023) updated the DJO12 data set and applied an estimator robust to mixed sample frequencies, finding that while temperature shocks exerted a temporary effect on income levels, they did not have a significant lasting effect on growth rates.

Newell *et al.* (2021) noted that there is no underlying theory to guide econometric model specification in this literature. Taking into account the arbitrary choices of which explanatory variables to include they identified over 800 possible model specifications. Using the Burke *et al.* (2015) data they used an estimation

method that accounts for model uncertainty and found that the model form preferred by Burke *et al.* (2015), which implied negative effects of warming on growth even in rich countries, is explicitly excluded by the optimal model selection algorithm. Overall they could not detect a temperature effect on GDP or GDP growth, and they estimated the 95 percent confidence interval for the impact on global growth as of 2100 even under the exaggerated RCP8.5 warming scenario spans –86 percent to +388 percent. In other words the net effect is likely positive but too uncertain to distinguish from zero.

Barker (2023) criticized the DJO12 assumption that countries can be grouped into fixed “poor” and “rich” categories based on their incomes many decades ago. He noted that many countries were once poor but became rich over time, and if this is considered the original temperature effects reported by Dell *et al.* became small and insignificant.

Berg *et al.* (2023) argued that countries shouldn’t be grouped into large Rich/Poor categories because they are too heterogeneous. They instead estimated country-specific temperature response coefficients then grouped countries with similar response coefficients into small panels. They separately estimated responses to global and idiosyncratic local temperature shocks to better identify the climate signal in weather data. Overall, they found countries experiencing negative effects of warming on growth outnumbered those experiencing positive effects, but only temporarily: eventually the effects reverse such that about twice as many countries experience a net positive growth effect. They also found global (as opposed to local) temperature changes are much more likely to benefit growth in poor countries than rich ones. In a simulation to 2100, even using the extreme RCP8.5 scenario, they computed the average global GDP loss would be only 1.9 percent compared to a scenario with no warming. That is, instead of the economy growing 400 percent it would grow 392%. The implication of Nordhaus’ earlier analysis is that trying to prevent the warming would lead to far less than 392 percent growth.

To summarize, economists consider climate a relatively unimportant factor in economic growth, a view echoed by the IPCC itself in the Fifth Assessment Report. Mainstream climate economics has recognized that CO₂-induced warming might have some negative economic effects but they are too small to justify aggressive abatement policy, and trying to “stop” or cap global warming even at levels well above the Paris target would be worse than doing nothing. An influential study in 2012 suggested that global warming would harm growth in poor countries but the finding has subsequently been found not to be robust (Tol 2024). Studies that take full account of modeling uncertainties either find no evidence of a negative effect on global growth from CO₂ emissions or find poor countries as likely to benefit from it as rich countries.

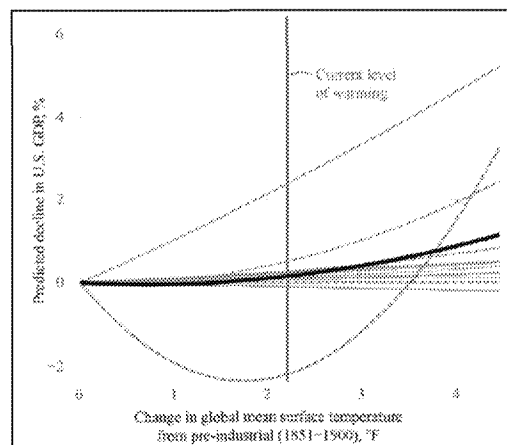


Figure 11.2: Decline in U.S. GDP per degree of warming. Source: CEA-OMB (2023)

The expectation that significant global warming would have a small impact on the U.S. economy was acknowledged quietly by the Biden Administration, even as the President was proclaiming a climate emergency. Figure 11.2, from a 2023 CEA-OMB report, shows the expected decrement in U.S. GDP as a function of temperature rise. The colored lines show the results of a dozen peer-reviewed published studies while the solid black line is their average. The figure could be summarized as “a few percent impact for a few degrees of warming”. Given that the economy’s annual growth rate is expected to be 1-2 percent, the impact of a warming globe on the U.S. GDP is indeed negligible.

11.2 Models of the Social Cost of Carbon

The Social Cost of Carbon (SCC) is a tool for quantifying the economic impact of carbon dioxide emissions, helping policymakers weigh the costs and benefits of climate policies. It estimates the damage caused by emitting one additional ton of CO₂, expressed in dollars. More formally, the SCC is the discounted present value of the current and future marginal loss of economic welfare due to an additional ton of CO₂ entering the atmosphere.

11.2.1 Estimating the SCC

Although the literature refers to “estimates” of the SCC, it is not estimated in the way other economic statistics are estimated. For instance, data on market transactions including prices and quantities can be used to estimate the current inflation rate or the growth rate of per capita real Gross Domestic Product, and there are well-understood uncertainties associated with these quantities. But there are no market data available to measure many, if not most, of the marginal damages or benefits believed to be associated with CO₂ emissions, so these need to be imputed using economic models.

For example, an influential component of some SCC calculations is the perceived social cost associated with a changed risk of future mortality due to extreme weather. There is no market in which people can directly attach a price to that risk. At best economists can try to infer such values by looking at transactions in related markets such as real estate or insurance, but isolating the component of price changes attributable to atmospheric CO₂ levels is very difficult.

Economists use IAMs to compute the SCC. Two of the best known are the Climate Framework for Uncertainty, Negotiation and Distribution (“FUND”, Tol 1997) and Nordhaus’ DICE. IAMs embed a “damage function” or set of functions relating ambient temperature to local economic conditions. The assumptions embedded in the damage function will largely determine the resulting SCC. IAMs also assume a long-term discount rate or, as in DICE, compute the optimal internal discount rate as part of the solution.

One approach to developing a damage function is to begin with estimates of the costs (or benefits) of warming in specific sectors in countries around the world and aggregate up to a global amount. This was the approach used in FUND. An alternative approach is to develop a simple equation that penalizes global income according to a simple quadratic function of the average global temperature. This approach was used in DICE. In the case of the FUND model many hundreds of parameters had to be selected, whereas in DICE only three were needed and were originally chosen to assign a predetermined penalty (1.2%) to global output as of 3°C warming, with a quadratic term implying that damages grow with the square of the global average temperature anomaly. Barrage and Nordhaus (2024) recently changed the parameters to increase the penalty at 3°C to 1.6%, and added a discrete additional 1.0 percent GDP penalty at 3°C warming to account for “tipping points” (discrete large-scale environmental changes triggered by crossing a warming

threshold) and a “judgmental adjustment” of 0.5 percent for excluded impacts at 3°C warming. Not surprisingly the newer version of DICE generates much higher SCC estimates than before.

The concepts of estimation and uncertainty do not readily apply to SCC calculations. By comparison, the question of whether precipitation has trended up or down along the South Carolina coast for the past hundred years can be addressed by collecting rainfall data, estimating a trend parameter and calculating statistical confidence intervals using methods with well-understood mathematical properties. In this way uncertainty can, in principle, be resolved by collecting more data and applying more accurate methods. But no amount of data collection can change the fact that many components of the SCC are unknown and rely on judgment and opinion based on knowledge of the underlying literature on the physical effects of climate change.

SCC calculations are thus best thought of as “if-then” statements: *if* the following assumptions hold, *then* the SCC is \$X per tonne. The list of ‘if’ statements includes the premise that the world’s climate and economy work according to the representation in the IAM. One way this might fail to hold is in the timing of warming. Every IAM assumes a value of the Equilibrium Climate Sensitivity (ECS), which controls the amount of warming that results from CO₂ emissions and can be freely varied for the purpose of generating a distribution of SCC values associated with uncertainties over ECS. But as Roe and Baumann (2013) pointed out, time-to-equilibrium increases with the square of ECS, so an upward adjustment of the ECS parameter without an appropriate slowing down of the adjustment process can yield distorted present value damage estimates. In particular, the upper tail of warming associated with some commonly-used ECS distributions is physically impossible for even a thousand years into the future (Roe and Baumann 2013) yet in an IAM would be realized within a couple of centuries. Failure to align ECS with time-to-equilibrium will lead to an overestimate of the SCC value.

11.2.2 Variations in the SCC

Every level of the IAM calculation includes assumptions, some more influential than others. Key assumptions include the following.

- The discount rate: Climate damages accrue over a long time horizon and costs a century or two in the future need to be discounted to the present. The higher the discount rate the smaller the present value of future damages and vice versa. The discount rate represents the opportunity cost of spending money today rather than investing it and then having more to spend tomorrow. Some economists have argued for the use of very low discount rates in SCC calculations, resulting in policy recommendations that favor relatively large immediate investments in CO₂ emission reductions. The downside is that other investments could potentially earn a larger rate of return for society.
- Equilibrium Climate Sensitivity: IAMs have customarily employed a value of 3.0°C or 3.1°C following the IPCC’s guidance. The most recent data-driven ECS values tend to be lower than this (see Chapter 4). Dayaratna *et al.* (2017, 2020, 2023) have shown that use of lower empirically-derived ECS values dramatically lower the resulting SCC estimate, even when low discount rates are used.
- Damage function coefficients: IAMs assume CO₂ and warming cause net harms that increase exponentially with temperature. More recently, IAMs have also incorporated effects from assumed potential climate tipping points. The FUND model took limited but explicit account of CO₂ fertilization effects in agriculture. Since the coefficients were selected prior the publication of the current evidence of global greening and the magnitude of benefits to crops from elevated CO₂ (see Chapters 2 and 9) the growth effects are likely understated. The DICE model (and others) did not

explicitly include any CO₂ fertilization benefit, except to the extent it was taken into account in the literature consulted when picking the damage function coefficients. The damage function in FUND contains a region in which low warming yields net benefits in many regions, a finding which is supported by econometric models of warming and growth (Berg *et al.* 2023) and econometric simulations of agricultural changes (McKittrick 2025). The DICE damage function, by construction, rules out net benefits at any warming level.

- Emission scenarios: IAMs generate SCC estimates that increase as the pre-existing concentration of CO₂ increases. Consequently the value of damages later in the century will be higher, depending on the assumed baseline emissions over the coming decades.
- Abatement costs: IAMs represent the cost to the economy of reducing CO₂ emissions. If CO₂ emission reductions are assumed to be inexpensive, then the model will conclude that the optimal policy should aim for deeper emission cuts and vice versa.

It is informative to ascertain whether SCC results are invariant to changes in some assumptions. But when different assumptions result in higher or lower SCC values, the change in the SCC value does not provide *prima facie* evidence about the validity of the assumptions. For example, in 2023 the U.S. Environmental Protection Agency raised its preferred SCC value about 5-fold over the estimates it had issued ten years earlier. This is not because new data had been collected or better mathematical methods had been invented, but because new assumptions had been used, and the validity of those assumptions was a separate question. One was that global agricultural damages were far higher than previously assumed, based on an analysis in Moore *et al.* (2017). But as discussed in Chapter 9, McKittrick (2025) showed that Moore *et al.* (2017) had used a database in which half the CO₂ change observations were missing. When as many of those observations as possible were recovered from underlying sources and the analysis was rerun, the projected crop yield losses disappeared and turned into gains at all warming levels. Hence the portion of the EPA's SCC revision attributable to agricultural yield losses was unwarranted.

11.2.3 Evidence for low SCC

Chapter 2 reviewed evidence on climate change and greening, and Chapter 9 looked at climate change and U.S. agriculture. Evidence shows that CO₂ fertilization has a stronger beneficial effect on agriculture than was known at the time that IAMs like DICE and FUND were parameterized. Haverd *et al.* (2020) found the observed CO₂ fertilization rate has been almost double what had been predicted by crop models. Dayaratna *et al.* (2023) used the updated empirical ECS distribution estimate of Lewis (2022), which assimilates modern instrumental and paleoclimatic temperature records, and allowed for a 30 percent gain in the CO₂ fertilization benefit in the FUND model, and found that, even at a low discount rate of 2.5 percent, the median SCC as of 2050 is only \$18.67, with a 24 percent probability of the true value being negative. At a five percent discount rate the median SCC value is negative until the mid-2040s and at 2050 was only \$0.37 with a 49 percent probability of being negative. Thus, under reasonable assumptions a mainstream IAM using updated scientific inputs yields evidence consistent with the SCC not being significantly greater than zero.

It should also be noted that the SCC is focused on the social costs of CO₂ emissions from fossil fuel use. It does not measure the private marginal benefits to consumers and society from the availability of fossil fuels. Public willingness to pay for fuels of all types indicates the value to society of reliable, abundant fossil energy. Tol (2017) estimates that the private benefit of carbon is large relative to the social cost. This can be illustrated by noting that the price of a gallon of gas indicates the marginal value to the consumer of the fuel. Suppose we assume a relatively high Social Cost of Carbon of, say, \$75 per tonne. Deflated by a MCPF⁴ value of 1.5 that would result in a carbon tax of \$50 per tonne, which equates to about 44 cents per

⁴ Marginal Cost of Public Funds: the optimal carbon tax rate is the SCC divided by the MCPF (Sandmo 1977).

gallon of gas (Lavelle, 2019). A pre-tax price of \$3.00 per gallon would imply the marginal social benefit of the fuel is nearly seven times the marginal social cost.

11.2.4 Tipping points

SCC calculations typically consider gradual impacts of a warming climate, such as slowly-melting glaciers and increasing average temperature. A driver of potentially high values of the Social Cost of Carbon (SCC) is the introduction into models of discrete catastrophic outcomes associated with abrupt changes (Dietz *et al.*, 2021). They are often referred to as “tipping points.”

The term “tipping point” mingles two different physical concepts that pose different research challenges. Many physical systems are inherently stable unless acted upon with sufficient external energy. For example, an ice cap might remain intact over a wide range of temperatures but once the temperature crosses the 0°C threshold it melts. Such discontinuities are ubiquitous in nature and require an external force. Whether the force needs to be large relative to the size of the system depends on the underlying stability of the system.

A different type of tipping point is called a *bifurcation* and arises from the study of the internal dynamics of nonlinear systems (Crawford, 1991). Many systems have been observed to have more than one equilibrium point and can move between them with minimal or no external influence. For example, a weather system might have two different equilibrium states: one calm and one with a tornado. A transition from one to another can happen either with no external force or with a minuscule change, such as a flap of the proverbial butterfly’s wings (Shen *et al.* 2014). The term “tipping point” is sometimes used to mean a bifurcation of this type and implies instability inherent in the system itself, which is not necessarily dependent on outside forces. It depends, instead, on parameters of the system taking values that support the emergence of bifurcations (Crawford, 1991).

The two different concepts imply different research questions. Regarding the first we are interested in whether components of the climate system are susceptible to abrupt discontinuities in response to sufficiently-large anthropogenic forcing. Regarding the second we are interested in whether the Earth’s climate system has inherent bifurcations that imply the possibility of abrupt transitions with or without external forcing.

Models have been developed that imply the second type is a possibility. Kypke *et al.* (2020) presented a simple climate model in which the GHG concentration is one of the parameters that controls the emergence of bifurcations of the Arctic climate to one with both cold and warm equilibria. If sufficiently high GHG forcing combined with a sufficiently high rate of ocean heat transport are imposed a bifurcation becomes possible. More generally GCMs have been observed to contain bifurcations and multiple equilibria (Brunetti and Ragon, 2023).

The possible existence of bifurcations in the Earth’s climate system implies abrupt transitions are possible, not just in response to large forcing but also to small perturbations. This places tipping points into the category of low-likelihood and potentially catastrophic events, such as large meteor strikes. A key question to ask is whether those kinds of tipping points can be predicted. Current research has not resolved that question (Dakos *et al.*, 2024) and indeed might not be able to since one implication of the “butterfly effect” is the existence of predictability boundaries of nonlinear systems (Palmer *et al.*, 2014). It is therefore not obvious how to incorporate such possibilities into SCC calculations. Small variations in assumptions will lead to arbitrarily large variations in the resulting SCC with no grounds for choosing among them. If such tipping points are possible the most appropriate stance for economic policy is to maximize resilience to any form of external catastrophe since it is unlikely we could predict it or prevent it from happening.

AR6 (WGI, Chapter 1) focuses mainly on the first type of tipping point, namely an abrupt change in response to external forcing. This is also the meaning associated with popular usage of the “tipping point” concept in discussions of climate change (see [[HYPERLINK "https://report-2023.global-tipping-](https://report-2023.global-tipping-)

points.org/what-is-a-tipping-point/"] for example). As summarized by AR6, there is evidence of abrupt change in the paleoclimate record, and some of these events have been interpreted as tipping points. Some projections with Earth System Models for example have produced tipping points such as Amazon forest dieback in response to specified values of CO₂ concentration or temperature increases.

The alarm surrounding climate tipping points is reflected by *The Global Tipping Points Report* that was launched at the COP28 on 6 December 2023 (Lenton *et al.*, 2023). It identifies more than 25 parts of the climate system said to constitute tipping points. What gets classified as a climate “tipping point” is a moving target. The most common examples in the literature and assessment reports include: Greenland ice sheet disintegration, West Antarctic ice sheet disintegration, summertime disappearance of Arctic sea ice, Amazon rainforest dieback, coral reef dieoff, thawing of permafrost and methane hydrates, Atlantic Meridional Overturning Circulation collapse, boreal forest shift, West African monsoon shift, and Indian Monsoon shift.

All such tipping points require a certain amount of system instability in order to exhibit an abrupt transition in response to warming. For this reason there seems to be very little to discriminate between a tipping event and natural climate variability. Natural climate variability has in the past produced shifts in the West African and Indian monsoons, Amazon forest and coral reef dieback, and disintegration of parts of the Greenland and West Antarctic ice sheets. These impacts can reverse on the decadal and century timescales associated with natural climate variability and ecosystem responses.

Some abrupt changes are potentially more consequential, including collapse of the West Antarctic ice sheet and collapse of the Atlantic Meridional Overturning Circulation (AMOC). AR6 WG1 Summary for Policy Makers states:

The Atlantic Meridional Overturning Circulation is very likely to weaken over the 21st century for all emission scenarios. While there is *high confidence* in the 21st century decline, there is only *low confidence* in the magnitude of the trend. There is *medium confidence* that there will not be an abrupt collapse before 2100. (C.3.4)

There is limited evidence for low-likelihood, high-impact outcomes (resulting from ice-sheet instability processes characterized by deep uncertainty and in some cases involving tipping points) that would strongly increase ice loss from the Antarctic Ice Sheet for centuries under high GHG emissions scenarios (e.g., SSP5-8.5). (B.5.2)

For the purpose of SCC calculations, the research question implied by this type of tipping point is whether such events have been observed in the past in climate conditions similar to what we currently experience or will in the near future. AR6 finds little evidence for impending collapse of the Atlantic Meridional Overturning Circulation or the West Antarctic ice sheet. It finds there is no tipping point associated with Arctic Sea ice (AR6 Technical Summary p. 76)

Dietz *et al.* incorporated several potential tipping points (abrupt changes) into an SCC model and found they added about 25 percent to the estimate, mainly associated with thawing permafrost and release of methane hydrates. However, the IPCC considers this scenario *very unlikely* (AR6 Technical Summary p. 107).

In summary, there might be unknown bifurcation tipping points that are associated with natural climate processes, but this possibility does not translate into specific guidance on the SCC. There are potential abrupt change points in the climate system in response to warming, although the IPCC assigns low probabilities to most, including the largest ones. When these have been considered, the result is only a modest increase in the SCC value in the 21st century.

11.2.5 Are there alternatives?

It is increasingly being argued that the SCC is too variable to be useful for policymakers. Cambridge Econometrics (Thoung, 2017) stated it's "time to kill it" due to uncertainties. The UK and EU no longer use SCC for policy appraisal, opting for "target-consistent" carbon pricing (UK Department for Energy Security and Net Zero 2022, Dunne 2017). However, the uncertainty of SCC estimates doesn't mean that other regulatory measures are inherently better or more efficient. Many emissions regulations (such as electric vehicle mandates, renewable energy mandates, energy efficiency regulations and bans on certain types of home appliances) cost far more per tonne of abatement than any mainstream SCC estimate, which is sufficient to establish that they fail a cost-benefit test.

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12 GLOBAL CLIMATE IMPACTS OF U.S. EMISSIONS POLICIES

Chapter Summary

U.S. policy actions are expected to have undetectably small direct impacts on the global climate and any effects will emerge only with long delays.

12.1 The scale problem

The emissions rates and atmospheric concentrations of criteria air contaminants are closely connected because their lifetimes are short and their concentrations are small; when local emissions are reduced the local pollution concentration drops rapidly, usually within a few days. But the global average CO₂ concentration behaves very differently, since emissions mix globally and the global carbon cycle is vast and slow. Any change in local CO₂ emissions today will have only a very small global effect, and only with a long delay.

Following the emission of a pulse (release) of CO₂ into the atmosphere only about 40 ± 15 percent of the emitted CO₂ will have been sequestered after twenty years. That fraction rises to 75 ± 10 percent after a thousand years, and the remainder will be gradually removed over the ensuing tens of thousands of years (Ciais *et al.*, 2013, pp.472-473). Consequently, any reduction in U.S. emissions would only modestly slow, but not prevent, the rise of global CO₂ concentration. And even if global emissions were to stop tomorrow, it would take decades or centuries to see a meaningful reduction in the global CO₂ concentration and hence human influences on the climate.

Reducing the atmospheric stock of CO₂ would require emissions to fall below the natural sequestration rate, assuming the entire increase is anthropogenic. Since that rate has been averaging about 50 percent of emissions in recent decades, a reduction of global emissions by 50 percent would (at least temporarily) halt the rise in atmospheric CO₂. The 1997 Kyoto Protocol proposed to cap industrial nations' CO₂ emissions at a modest five percent below 1990 levels by the year 2012. Even though this policy was too difficult for most nations to implement, full compliance would not have substantially reduced atmospheric CO₂ levels. It would only have slightly slowed CO₂ growth, reaching the projected year 2100 level in 2105 instead (Wigley, 1998). Lomborg (2016) estimated that full compliance with the Paris Accord would not stop warming, it would only prevent about 0.1C warming and delay hitting the baseline year 2100 temperature levels by about a decade.

Thus, in contrast with conventional air pollution control, even drastic local actions will have negligible local effects, and only with a long delay. The practice of referring to unilateral U.S. reductions as “combatting climate change” or “taking action on climate” therefore reflects a profound misunderstanding of the scale of the issue.

12.2 Case study: U.S. motor vehicle emissions

The scale problem can be illustrated with reference to U.S. motor vehicles. The EPA's 2009 Endangerment Finding focused on CO₂ emissions from cars and light-duty trucks in the U.S. because Section 202(a) of Clean Air Act mandates the EPA to set emissions standards for motor vehicles if pollutants are found to endanger public health or welfare. The 2009 Endangerment Finding therefore obligated the EPA to regulate emissions from new motor vehicles, ostensibly to reduce or eliminate climate-related harms to the U.S. public.

Two questions that naturally arise are: (1) How large a reduction in CO₂ emissions would result from such regulation? and (2) What would be the climate impact of such regulation?

The first question can be addressed by comparing U.S. vehicle-based CO₂ emissions to the global total. The second question can be addressed by using the fact that the reduction in global warming would be, according to the models relied upon by the EPA, proportional to the reduction in global emissions, keeping in mind that the change in the CO₂ content of the atmosphere in any given year is the result of total global CO₂ emissions, not just U.S. emissions.

In 2022, the emissions from U.S. cars and light duty trucks totaled 1.05 billion metric tons of carbon dioxide (GtCO₂, EPA 2024). Meanwhile global CO₂ emissions from energy use totaled 34.6 GtCO₂ (Energy Institute 2024). Hence U.S. cars and light trucks account for only 3.0 percent of global energy-related CO₂ emissions. To a first approximation we can say that even eliminating *all* U.S. vehicle-based emissions would retard the accumulation of CO₂ in the atmosphere by a year or two over a century.

It would also reduce the overall warming trend by at most about 3 percent. For the period 1979-2023, which has the most extensive global coverage of a variety of weather data types, warming trends are determined to a precision of about ± 15 percent, so the impact of reducing the rate of global warming by eliminating U.S. vehicle CO₂ emissions would be far below the limits of measurability. Given that global-average temperature is the most direct climate change metric, impacts on any secondary climate metrics (e.g. severe weather, floods, drought, etc.) from reducing U.S. vehicle CO₂ emissions would be even less measurable.

Consequently, in contrast to the case of local air contaminants like particulates and ozone, even the most aggressive regulatory actions on GHG emissions from U.S. vehicles cannot be expected to remediate alleged climate dangers to the U.S. public on any measurable scale.

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GLOSSARY

2XCO₂: Doubling of the atmospheric carbon dioxide (CO₂), a commonly used baseline for addressing climate change. In 2025, atmospheric CO₂ concentrations were approximately 50 percent of the way to 2XCO₂ compared to 1800.

ABATEMENT: In the context of environmental policy, the reduction of emissions.

ACE (Accumulated Cyclone Energy): A statistical measure of a tropical cyclone's energy accumulated over its lifetime, calculated from a summation of the square of maximum sustained wind speeds.

ACIDIFICATION: A commonly used term for the reduction in ocean pH from more alkaline to less alkaline values.

ACRIM (Active Cavity Radiometer Irradiance Monitor) GAP: the gap in total solar irradiance data between 1989 and 1991 caused by a delay in the launch of a satellite-borne monitor in time to overlap and intercalibrate with the prior system.

AEROSOLS: Tiny solid or liquid particles suspended in the air.

ALKALINE: Having a chemical pH greater than 7.0. The "opposite" of acidic, which means a pH less than 7.0.

AMBIENT: Relating to the immediate surroundings of something.

AMO: Atlantic Multidecadal Oscillation, a 60-year cycle of sea surface temperatures in the North Atlantic.

ANTHROPOGENIC: Created by humans.

AR (Assessment Reports): Periodic assessments published by the Intergovernmental Panel on Climate Change evaluating current knowledge of the climate system, global change, and associated issues. An AR now consists of reports from three working groups, the first (WG1) concerned with physical science and the others concerned with societal impacts and mitigation strategies. The most recent was AR6.

ATTRIBUTION: Assertion of a causal relation, chiefly from anthropogenic GHG emissions to observed climate patterns.

AUTOCORRELATION: The phenomenon in which the current value of a random time series variable is correlated with its value in a previous time period.

BIOMASS: The total quantity or weight of organic matter in a given area or volume.

CARBON: The 6th element in the periodic table. Popularly used as a shorthand term for carbon dioxide.

CMIP (Coupled Model Intercomparison Project): The Intergovernmental Panel on Climate Change's organized effort to compare many different global ocean-atmosphere climate models when they are all forced with the same emissions scenario(s). The two most recent were the CMIP5 and CMIP6 for, respectively, AR5 and AR6.

CLIMATE CHANGE: A change in climate due to human influences

CONUS: The contiguous 48 U.S. states.

DICE (Dynamic Integrated model of the Climate and Economy): An Integrated Assessment Model developed by William Nordhaus in 1993.

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DISCOUNT RATE: The interest rate used to convert future costs and benefits into their present value. Varies inversely with present value so a high discount rate reduces the present value of future costs and benefits.

DRY ICE: Frozen carbon dioxide.

ECONOMETRICS: The branch of economics concerned with the use of statistical methods (chiefly multiple regression) to analyse economic systems.

ECS (Equilibrium Climate Sensitivity): The total amount of eventual global-average surface warming in response to a hypothetical doubling of the atmospheric CO₂ concentration compared to pre-Industrial levels.

EL NIÑO: The warm phase of the El Niño – Southern Oscillation, involving weaker upwelling of cold water off the South American coast and weaker Pacific trade winds.

EMISSIONS SCENARIO: An assumed scenario of future greenhouse gas emissions (or resulting atmospheric concentrations) based upon assumptions regarding global economic activity, the prevalence of fossil fuel use, and (sometimes) global carbon cycle model estimates of the rate of CO₂ uptake by land and ocean.

ENDANGERMENT FINDING (EF): The 2009 finding by the Administrator of the Environmental Protection Agency that emissions of well-mixed greenhouse gases (primarily CO₂) endanger human health and welfare.

ENSO: El Niño – Southern Oscillation, a prominent natural climate fluctuation involving year-to-year variations in the upwelling of cold water off the equatorial coast of South America and associated strengthening or weakening of trade winds across the Pacific Ocean.

FACE: (Free-Air CO₂ Enrichment) Large-scale open-air studies that expose plants to elevated CO₂ levels, simulating future climate scenarios.

FOSSIL FUEL: A natural hydrocarbon fuel such as coal, oil, or natural gas formed in the geological past from the remains of living organisms.

FUND (Framework for Uncertainty, Negotiation, and Distribution): An Integrated Assessment Model developed by Richard Tol in 1997.

GBR (Great Barrier Reef): The world's largest coral reef ecosystem located off the northeast coast of Australia.

GDP (Gross Domestic Product): Total market value of all the final goods and services produced within a country's border, usually over one year.

GHCN (Global Historical Climatology Network): A time-varying number of globally distributed weather stations making hourly, daily, or monthly measurements of precipitation, temperature and sometimes snowfall and snow depth.

GLOBAL GREENING: The satellite-observed increase in greenness in most vegetated land areas observed since the early 1980s.

GNP (Gross National Product): Total market value of all the final goods and services produced by the citizens of a country, including exports and imports, usually over one year.

GDP (Gross National Product): Total market value of all the domestically-produced final goods and services produced by the citizens of a country, usually over one year.

GREENHOUSE EFFECT (GHE): The tendency for any planetary atmosphere containing greenhouse gases to be warmer in the lowest layers than if those gases did not exist.

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GREENHOUSE GAS (GHG): An atmospheric gas that absorbs and emits infrared radiation, especially water vapor, CO₂, and methane.

HURST PHENOMENON: See long term persistence.

HYDROLOGY: The study of the movement of water, especially on land and in the atmosphere.

IAM (Integrated Assessment Model): A computerized tool that combines economics, climate science, and social sciences to provide quantitative relationships between human and Earth systems, helping to inform policy decisions.

IEA (International Energy Agency): A Paris-based intergovernmental organization, established in 1974, that provides policy recommendations, analysis and data on the global energy sector.

IPCC (Intergovernmental Panel on Climate Change): A panel of international experts and governmental representatives established by the United Nations and World Meteorological Organization in 1988 to provide governments at all levels with scientific information that they can use to develop climate policies.

IR (Infrared): Heat radiation that all solid objects and greenhouse gases emit by virtue of their temperature.

LA NIÑA: The cool phase of ENSO, involving stronger upwelling of cold water off the South American coast and stronger Pacific trade winds.

LONG TERM PERSISTENCE (LTP, or HURST PHENOMENON): A form of autocorrelation in which random fluctuations induce long term but not permanent changes in the mean value of a series, which shows up in some climate data series as cyclical or trending patterns on short time scales.

MAIZE: Corn.

MODEL: A computer code or collection of codes which quantify our knowledge of various processes and the interactions between those processes.

MORTALITY: Death or number of deaths, usually expressed for a specific population and period.

NCA (National Climate Assessment): A periodic report mandated by the Global Change Research Act of 1990 summarizing the impact of climate change on the United States. There have been five NCA reports: NCA1 (2000), NCA2 (2009), NCA3 (2014), NCA4 (2017-18), and NCA5 (2023).

OECD (Organization for Economic Cooperation and Development): A group of 38 member countries that have a democratic system of government and free economic systems.

PALEOCLIMATE: Climate of the distant past, before modern weather instruments were widely used (typically pre-1850), necessitating the use of “proxy” measurements such as tree rings, ice core, pollen records, etc.

PDI (Power Dissipation Index): A statistical estimate of the accumulated destructive potential of a tropical cyclone over time, calculated as a summation of the cube of the maximum sustained wind speeds.

PDO: Pacific Decadal Oscillation, a recurring ocean-atmosphere interaction centered over the Pacific basin associated with predominantly warm or cool patterns that influence the global climate.

PHOTOSYNTHESIS: The process by which plants grow, requiring carbon dioxide, water, and light at sufficiently warm temperatures (10 to 35 deg. C for optimal function).

PPM: Parts per million, a measure of the concentration of a substance in (for example) the atmosphere

PRE-INDUSTRIAL: The near-term historical period before which human greenhouse gas emissions were considered significant, usually assumed to be the mid-1700s.

RADIATIVE FEEDBACK: A change in the balance between absorbed solar radiation and emitted infrared (IR) radiation, referenced to the top of the atmosphere, caused by changes in surface temperature.

RADIATIVE FORCING: The change in the balance between absorbed solar radiation and emitted infrared (IR) radiation, referenced to the top of the atmosphere, caused by changes in greenhouse gases, anthropogenic aerosols, volcanoes, etc.

RCP (Representative Concentration Pathway) SCENARIOS: Different scenarios for future greenhouse gas emissions and their impact on the atmosphere, introduced in the IPCC's Fifth Assessment Report (AR5). RCP scenarios have a number label (e.g. RCP6.0) indicating how much radiative forcing they assume in 2100 relative to pre-Industrial times, in Watts per square meter. RCP4.5 and RCP6.0 are intermediate scenarios, whereas RCP8.5 is an extreme scenario with very large future emissions of greenhouse gases.

REGRESSION: A statistical method for selecting the coefficients of a linear equation to explain the behaviour of a dependent variable in terms of variations in one or more explanatory variables.

RICARDIAN ANALYSIS: A method used in environmental economics named after David Ricardo to estimate the economic impacts of climate change, particularly on agriculture, by making use of the hypothesis that expected changes in the return to future agricultural activity will be capitalized into current land values.

SCC (Social Cost of Carbon): An estimate of the net economic damages, measured in present-value dollars, caused by emitting one additional ton of CO₂ into the atmosphere.

SSP (Shared Socioeconomic Pathway) SCENARIOS: Different scenarios for future greenhouse gas emissions and their impact on the atmosphere, introduced in the IPCC's Sixth Assessment Report (AR6). Some of the SSP scenarios are analogous to the older RCP scenarios, for purposes of continuity and comparison to earlier climate model assessments.

TC: Tropical cyclone.

TMAX: The daily maximum surface air temperature.

TMIN: The daily minimum surface air temperature.

TOA: Top of atmosphere.

TOXICOLOGY: The study of the adverse effects of chemical substances on living organisms.

TREND: A coefficient, usually estimated using linear regression, representing the slope of a line of best fit through a time series of data, indicating any tendency for the series mean to move up or down over time.

TSI (Total Solar Irradiance): The measure of the total amount of incident solar radiation per unit area, including all wavelengths of electromagnetic radiation, that reaches the Earth's atmosphere. Typically shown in units of watts per square meter (W/m²).

UHI (Urban Heat Island): The tendency for inhabited areas to be warmer than their rural surroundings, due to the replacement of natural land and vegetation with roads, parking lots, buildings, and waste heat sources.

USHCN: The U.S. Historical Climatology Network, consisting of quality-controlled temperature and precipitation data from 1,218 surface weather stations.

METADATA FOR FIGURES AND TABLES

Figure 2.1 Screen shot from Figure 3 of Zhu *et al.* 2016 [HYPERLINK "<https://www.nature.com/articles/nclimate3004>"]

Figure 2.2 Data replotted from Berner 2006 Figure 18, with annotation in red added by authors [HYPERLINK "<https://www.sciencedirect.com/science/article/pii/S0016703706002031>"]

Figure 2.3 Screen shot from Gerhart and Ward (2010) Figure 2 [HYPERLINK "<https://nph.onlinelibrary.wiley.com/doi/pdf/10.1111/j.1469-8137.2010.03441.x>"]

Figure 2.4 Screenshot from Copernicus Marine Services [HYPERLINK "https://data.marine.copernicus.eu/product/GLOBAL_OMI_HEALTH_carbon_ph_area_averaged/description"]

Figure 2.5 Screen shot from AIMS (2023) Figures 2-4 [HYPERLINK "https://www.aims.gov.au/sites/default/files/2022-08/AIMS_LTMP_Report_on%20GBR_coral_status_2021_2022_040822F3.pdf"]

Figure 3.1.1 Screen shot of AR6 WGI Ch2 Fig 10

Figure 3.1.2 Screen shot of AR6 WGI Ch7 Fig 7-6

Figure 3.1.3 Author created figure. CO2 data source: [HYPERLINK "<https://gml.noaa.gov/ccgg/trends/index.html>"]

Figure 3.2.1 Screen shot from Hausfather *et al.* (2019) Figure S4 [HYPERLINK "<https://doi.org/10.1029/2019GL085378>"]

Figure 3.2.2 Author created figure. Data source: Freidlingstein *et al.* (2024) [HYPERLINK "<https://essd.copernicus.org/preprints/essd-2024-519>"]
Data source: [HYPERLINK "<https://globalcarbonbudget.org/download/1442/>"]

Figure 3.2.3 Author-created using OLS trends, Data source: [HYPERLINK "<https://globalcarbonbudget.org/download/1442/>"]

Figure 3.2.4 Author-created using OLS trends, Data source: [HYPERLINK "<https://globalcarbonbudget.org/download/1442/>"]

Figure 4.1 Screen shot from Scafetta (2021) Figure 1 [HYPERLINK "<https://doi.org/10.3390/cli9110161>"]

Figure 5.1 Data source: [HYPERLINK "<https://climexp.knmi.nl/start.cgi>"]. Author-created figure. One ensemble member per model, yearly average temperatures are computed from the global monthly raw temperatures; 'range' is the warmest model temperature minus the coolest model temperature in each year; standard deviation computed across all 33 models' raw temperatures in each year.

Figure 5.2 Screen shot from Scafetta (2023) Figure 2 [HYPERLINK "<https://doi.org/10.1007/s00382-022-06493-w>"]

Figure 5.3 Author created figure. Data source: [HYPERLINK "<https://climexp.knmi.nl/start.cgi>"]. One SSP370 ensemble member per model. Linear OLS trends computed from global monthly temperature anomalies relative to 1981-2010.

Figure 5.4 Author created figure, based on Figure 3 of McKittrick and Christy (2020) [HYPERLINK "<https://doi.org/10.1029/2020EA001281>"]. Updated using same methods and data extended from 2014 to 2024, see McKittrick and Christy (2025) in chapter references.

Figure 5.6 Screen shot of IPCC AR5 Figure 10.SM1, with author annotations

Figure 5.7 Author created figure following Christy and McNider (2017) [HYPERLINK "<https://doi.org/10.1007/s13143-017-0070-z>"] with updated data available at [HYPERLINK "https://www.nsstc.uah.edu/data/cmip6/ERA5_REANALYSES/"] and [HYPERLINK "https://www.nsstc.uah.edu/data/cmip6/JRA3Q_REANALYSES/"] compiled here [HYPERLINK "https://www.nsstc.uah.edu/data/cmip6/CWG25_Fig_5.7_Table_250627.xlsx"]

Figure 5.8 Screen shot from [HYPERLINK "https://climate.rutgers.edu/snowcover/chart_seasonal.php?ui_set=nhland&ui_season=1"] (accessed May 27, 2025)

Figure 5.9 Screen shot from Rugenstein and Hakuba (2023) Figure 1 [HYPERLINK "<https://doi.org/10.1029/2022GL101802>"]

Figure 5.10 Author created figure. Data source [HYPERLINK "<https://climexp.knmi.nl/start.cgi>"]. OLS linear trends. One SSP370 ensemble member per model, averaged surface air temperatures for June through August over 31N to 49N latitude, 84W to 102W longitude. Observed temperature data for same months from NOAA/NCEI at <https://www.ncei.noaa.gov/access/monitoring/climate-at-a-glance/statewide/time-series> for the 12 Corn Belt states averaged together.

Figure 6.1.1 Author replot of Figure 10 from Koutsoyiannis (2013) [HYPERLINK "<http://dx.doi.org/10.1080/02626667.2013.804626>"]

Figure 6.2.1 Screen shot of figure from Maue (2025) [HYPERLINK "<https://climatlas.com/tropical/>"]

Figure 6.2.2 Author created figure. Data from National Hurricane Center (2024) [HYPERLINK "<https://www.nhc.noaa.gov/climo/>"]

Figure 6.2.3 Author created figure. Data from NOAA HRD [HYPERLINK "https://www.aoml.noaa.gov/hrd/hurdat/All_U.S._Hurricanes.html" \t "_new"]

Table 6.2.1 Author created table. Data from NOAA HRD [HYPERLINK "https://www.aoml.noaa.gov/hrd/hurdat/most_intense.html" \t "_new"]

Figure 6.3.1 Screen shot from NCA4 Figure 6.4

Figure 6.3.2 Author created figure. Data source: [HYPERLINK "https://www.nsstc.uah.edu/data/ushcn_jrc/"]

Figure 6.3.3 Author created figure. Data source: [HYPERLINK "https://www.nsstc.uah.edu/data/ushcn_jrc/"]

Figure 6.3.4 Author created figure. Data source: [HYPERLINK "https://www.nsstc.uah.edu/data/ushcn_jrc/"]

Figure 6.3.5 Author created figure Data source: [HYPERLINK "https://www.nsstc.uah.edu/data/ushcn_jrc/"]

Figure 6.3.6 Author created figure. Data source: [HYPERLINK "https://www.nsstc.uah.edu/data/ushcn_jrc/"]

Figure 6.3.7 Screen shot from [HYPERLINK "https://www.globalchange.gov/indicators/heat-waves"]

Box tables. Author created table. Data from McKittrick and Christy (2025)

Figure 6.4.1 Author created figure. Data from McKittrick and Christy (2025)

Figure 6.4.2 Author created figure. Data from McKittrick and Christy (2025)

Figure 6.4.3 Author created figure. Data from McKittrick and Christy (2025)

Figure 6.4.4 Author created figure. Data from McKittrick and Christy (2025)

Figure 6.4.5 Author created figure. Data from McKittrick and Christy (2025)

Figure 6.5.1 Author created figure. Data source: [HYPERLINK "https://www.spc.noaa.gov/wcm/data/1950-2024_actual_tornadoes.csv"]

Figure 6.7.1 Author created figure. Data source: [HYPERLINK "https://www.ncei.noaa.gov/access/monitoring/uspa/wet-dry/0"] accessed June 16, 2025. OLS trend line added.

Figure 6.8.1 Screen shot from Lizundia-Loiola *et al.* (2021) Figure 12.

Figure 6.8.2 Screenshot from Marlon *et al.* (2012) Figure 2 panel C. [HYPERLINK "https://www.pnas.org/doi/pdf/10.1073/pnas.1112839109"]

Figure 6.8.3 Author created Figure. Data 1926 to 2016: [HYPERLINK "https://web.archive.org/web/20200212033452/https://www.nifc.gov/fireInfo/fireInfo_stats_totalFires.html"]. Post-2017 [HYPERLINK "https://www.nifc.gov/fire-information/statistics/wildfires"] Accessed June 16, 2025.

Figure 7.1 Screenshot from [HYPERLINK "https://tidesandcurrents.noaa.gov/sltrends/"]

Table 7.1 Author created table. From Section 6.2 of [HYPERLINK "https://judithcurry.com/wp-content/uploads/2018/11/special-report-sea-level-rise-3.pdf"]

Figure 7.2 Screenshot from [HYPERLINK "https://tidesandcurrents.noaa.gov/sltrends/sltrends_station.shtml?id=9414290"] (downloaded 4/22/25)

Figure 7.3 Screenshot from

[HYPERLINK "https://tidesandcurrents.noaa.gov/sltrends/sltrends_station.shtml?id=8771450"]
(downloaded 4/22/2025)

Figure 7.4 Screenshot from

[HYPERLINK "https://tidesandcurrents.noaa.gov/sltrends/sltrends_station.shtml?id=8761724"]
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Figure 7.5 Screenshot from

[HYPERLINK "https://tidesandcurrents.noaa.gov/sltrends/sltrends_station.shtml?id=8518750"]
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Figure 7.6 Author created figure. Data from [HYPERLINK

"https://tidesandcurrents.noaa.gov/sltrends/sltrends_station.shtml?id=8518750"]

Figure 8.1 Screenshot from

[HYPERLINK "<https://crudata.uea.ac.uk/cru/data/temperature/HadCRUT5.0Analysis.pdf>"]

Figure 8.2 Screenshot from Hansen and Karecha (2025) Figure 1

[HYPERLINK "<https://www.columbia.edu/~jeh1/mailings/2025/CloudFeedback.13May2025.pdf>"]

Table 8.1 Screenshot of column 1 of Table 12.12 IPCC AR6 Working Group I report

Figure 9.1 Screenshot From Taylor and Schlenker (2021) Figure 1

[HYPERLINK "<https://www.nber.org/papers/w29320>"]

Figure 9.2 Screenshot from McKittrick (2025) Figure 1

[HYPERLINK "<https://www.nature.com/articles/s41598-025-90254-2>"]

Figure 10.1 Screenshot from Pielke (2024) [HYPERLINK "<https://www.nature.com/articles/s44304-024-00011-0>"] Figure 3.

Figure 10.2 Screenshot from Gasparini *et al.* (2015) Figure 2

[HYPERLINK "[http://dx.doi.org/10.1016/S0140-6736\(14\)62114-0](http://dx.doi.org/10.1016/S0140-6736(14)62114-0)"]

Figure 11.1 Screen shot from Pielke Jr (2023)

[HYPERLINK "<https://rogerpielkejr.substack.com/p/global-disaster-losses1990-2023>"]

Figure 11.2 Screenshot from CEA-OMB (2023) [HYPERLINK

"<https://bidenwhitehouse.archives.gov/wp-content/uploads/2023/03/CEA-OMB-White-Paper.pdf>"]
Figure 1.

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The Honorable Steven E. Koonin, Ph.D. is the Edward Teller Senior Fellow at Stanford's Hoover Institution. He has served as Under Secretary for Science at the Department of Energy (2009-2011), as Chief Scientist for British Petroleum (2004-2009), and as a professor at Caltech (1975 – 2004, the last nine years as Vice President and Provost). Koonin is a member of the National Academy of Sciences and the JASON group of government advisors, as well as a Governor of Lawrence Livermore National Laboratory, having served in similar capacities for the Los Alamos, Sandia, Brookhaven, and Argonne National Laboratories. Other prior roles include a University Professor at New York University, a non-resident Senior Fellow at the American Enterprise Institute, and a Trustee of the Institute for Defense Analyses. Koonin has a BS in Physics from Caltech and a PhD in Theoretical Physics from Massachusetts Institute of Technology. He authored the 2021 bestseller *Unsettled: What Climate Science Tells Us, What It Doesn't, and Why It Matters*, the 1985 textbook *Computational Physics*, and some 200 peer-reviewed papers in physics and astrophysics, scientific computation, energy technology and policy, and climate science.

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National Academy of Sciences report on Surface Temperature Reconstructions and has testified before the U.S. Congress and committees of the Canadian House of Commons and Senate.

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Impacts of Carbon Dioxide Emissions on the U.S. Climate

Climate Working Group
United States Department of Energy
July 15, 2025

Impacts of Carbon Dioxide Emissions on the U.S. Climate

Report to U.S. Energy Secretary Christopher Wright

July 15, 2025

Climate Working Group:

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Suggested citation:

Climate Working Group (2025) Impacts of Carbon Dioxide Emissions on the U.S. Climate. Washington DC: Department of Energy, July 15, 2025

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SECRETARY'S FOREWORD

Energy, Integrity, and the Power of Human Potential

Over my lifetime, I've had the privilege of working as an energy entrepreneur across a range of fields—nuclear, geothermal, natural gas, and more—and I now serve as Secretary of Energy under President Donald Trump. But above all, I'm a physical scientist who sees modern energy as nothing short of miraculous. It powers every aspect of modern life, drives every industry, and has made America an energy powerhouse with the ability to fuel global progress.

The rise of human flourishing over the past two centuries is a story worth celebrating. Yet we are told—relentlessly—that the very energy systems that enabled this progress now pose an existential threat. Hydrocarbon-based fuels, the argument goes, must be rapidly abandoned or else we risk planetary ruin.

That view demands scrutiny. That's why I commissioned this report: to encourage a more thoughtful and science-based conversation about climate change and energy. With my technical background, I've reviewed reports from the Intergovernmental Panel on Climate Change, the U.S. government's assessments, and the academic literature. I've also engaged with many climate scientists, including the authors of this report.

What I've found is that media coverage often distorts the science. Many people—even well-meaning ones—walk away with a view of climate change that is exaggerated or incomplete. To provide clarity, I asked a diverse team of independent experts to summarize the current state of climate science, with a focus on how it relates to the United States.

I didn't select these authors because we always agree—far from it. In fact, they may not always agree with each other. But I chose them for their rigor, honesty, and willingness to elevate the debate. I exerted no control over their conclusions. What you'll read are their words, drawn from the best available data and scientific assessments.

I've reviewed the report carefully and challenged its findings. I believe it faithfully represents the state of climate science today. Still, many readers may be surprised by its conclusions—which differ in important ways from the mainstream narrative. That's a sign of how far the public conversation has drifted from the science itself.

To correct course, we need open, respectful, and informed debate. That's why I'm inviting public comment on this report. Honest scrutiny and scientific transparency should be at the heart of our policymaking.

Climate change is real, and it deserves attention. But it is not the greatest threat facing humanity. That distinction belongs to global energy poverty. As someone who values data, I know that improving the human condition depends on expanding access to reliable, affordable energy. Climate change is a challenge—not a catastrophe. But misguided policies based on fear rather than facts could truly endanger human well-being.

We stand at the threshold of a new era of energy leadership. If we empower innovation rather than restrain it, America can lead the world in providing cleaner, more abundant energy—lifting billions out of poverty, strengthening our economy, and improving our environment along the way.

EXECUTIVE SUMMARY

This report reviews scientific certainties and uncertainties in how anthropogenic carbon dioxide (CO₂) emissions have affected, or will affect, the Nation's climate, extreme weather events, and selected metrics of societal well-being. Those emissions are increasing the concentration of CO₂ in the atmosphere through a complex and variable carbon cycle, where some portion of the additional CO₂ persists in the atmosphere for centuries.

Elevated concentrations of CO₂ directly enhance plant growth, globally “greening” the planet and increasing agricultural productivity. They also make the oceans less alkaline (lower the pH) by an amount within the range of natural variability. That is possibly detrimental to coral reefs, although the recent rebound of the Great Barrier Reef suggests otherwise.

Carbon dioxide also acts as a greenhouse gas, exerting a warming influence on climate and weather. Scenarios of future CO₂ emissions have tended to overestimate observed levels.

The world's several dozen detailed climate models offer little guidance on how the climate responds to elevated CO₂, with the average surface warming under a doubling of the CO₂ concentration ranging from 1.8°C to 5.7°C, a larger range than data driven estimates that suggest a response in the range 1.8°C to 2.7°C.

Detailed climate models generally run “hot” in their description of the climate of the past few decades - too much warming at the surface and too much amplification of warming in the lower- and mid-troposphere. They also produce too much stratospheric cooling, too much snow loss, and too much warming in the U.S. Corn Belt. The combination of overly sensitive models and extreme scenarios for future emissions has exaggerated projections of future warming.

Most extreme weather events do not show long-term trends. Claims of increased frequency or intensity of hurricanes, tornadoes, floods, and droughts are not supported by historical data. Additionally, forest management practices are often overlooked in assessing wildfire activity. Global sea level has risen approximately 8 inches since 1900, but there are significant regional variations in tide gauge data driven primarily by local land subsidence; U.S. tide gauge measurements in aggregate show no obvious acceleration in sea level rise beyond the historical average rate.

The attribution of particular climate changes or extreme weather events to human CO₂ emissions is challenged by natural climate variability, data limitations, and inherent model deficiencies. Moreover, solar activity's contribution to the late 20th century warming might be underestimated.

Both models and experience suggest that CO₂-induced warming might be less damaging economically than commonly believed, and excessively aggressive mitigation policies could prove more detrimental than beneficial. Social Cost of Carbon estimates, which attempt to quantify the economic damage of CO₂ emissions, are highly sensitive to their underlying assumptions and so provide limited independent information.

The report supports a more nuanced and evidence-based approach for informing climate policy. The risks and benefits of a climate changing under both natural and human influences must be weighed against the costs, efficacy, and collateral impacts of any “climate action”, considering the nation's need for reliable and affordable energy with minimal local pollution. Beyond continuing precise, un-interrupted observations of the global climate system, it will be important to make realistic assumptions about future emissions, re-evaluate climate models to address biases and uncertainties, and clearly acknowledge the limitations of extreme event attribution studies. An approach that acknowledges both the potential risks and benefits of CO₂, rather than relying on flawed models and extreme scenarios, is essential for informed and effective decision-making.

PREFACE

This document originated in late March 2025 when Secretary Wright assembled an independent group to write a report on issues in climate science relevant for energy policymaking, with particular focus on the question of whether carbon dioxide emissions endanger the U.S. public. We agreed to undertake the work on the condition that there would be no editorial oversight by the Secretary, the Department of Energy, or any other government personnel. This condition has been honored throughout the process and the writing team has worked with full independence.

The group began working in early April with a deadline of May 28. The short timeline and the technical nature of the material meant that we could not comprehensively review all topics. Rather, we chose to focus on topics that are treated by a serious, established academic literature; that are relevant to our charge; that are downplayed in, or absent from, recent assessment reports; and that are within our competence.

While the report is intended to be accessible to non-experts, we have omitted some introductory or explanatory material that can easily be accessed elsewhere. Nor have we attempted to survey entire literatures related to the topics covered. We have focused as much as possible on literature published since 2020 and referenced previous IPCC and NCA assessment reports. We have also used data through 2024 where possible.

The writing team is grateful to Secretary Wright for the opportunity to prepare this report and for his support of independent scientific assessment and open scientific debate.

John Christy, Ph.D.

Judith Curry, Ph.D.

Steven Koonin, Ph.D.

Ross McKittrick, Ph.D.

Roy Spencer, Ph.D.

**PART I: DIRECT HUMAN INFLUENCE ON ECOSYSTEMS AND THE
CLIMATE**

1 CARBON DIOXIDE AS A POLLUTANT

Chapter summary:

Carbon dioxide (CO₂) differs in many ways from the so-called Criteria Air Pollutants. It does not affect local air quality and has no human toxicological implications at ambient levels. It is an issue of concern because of its potential effects on the global climate, a complex issue that occupies much of this report.

The Clean Air Act of 1970 defined six so-called Criteria Air Contaminants subject to regulation (EPA): particulate matter, ground-level ozone, sulfur dioxide, nitrogen dioxide, lead, and carbon monoxide. In 2007, the Supreme Court ruled that greenhouse gases (CO₂ among them) were also “pollutants” subject to regulation under Clean Air Act (*Mass. v. EPA*, 2007). While the definition of “pollutant” is ultimately a legal matter, there are important scientific distinctions between CO₂ and the Criteria Air Contaminants. The latter are subject to regulatory control because they cause local problems depending on concentrations that include nuisances (odor, visibility), damage to plants, and, at high enough exposure levels, toxicological effects in humans. In contrast, CO₂ is odorless, does not affect visibility and has no toxicological effects at ambient levels. It is a naturally occurring part of the atmosphere and a key component of human and plant respiration. CO₂ is essential for plant photosynthesis and higher levels are beneficial for vegetation. In these aspects, CO₂ is similar to water vapor.

Ambient outdoor air today contains about 430 parts per million (ppm) CO₂, increasing at about 2 ppm per year. The U.S. Occupational Safety and Health Administration issues guidelines for indoor workplaces in which elevated CO₂ might be encountered, such as where dry ice is used. The Permissible Exposure Limit is 5,000 ppm over 8 hours (OSHA, 2024). Allen *et al.* (2015) reported evidence of diminished performance on some cognitive tasks among workers in office cubicles when exposed to CO₂ levels above 1,000-1,500 ppm. These levels are far larger than any plausible ambient outdoor value through the end of the 22nd century.

The growing amount of CO₂ in the atmosphere directly influences the earth system by promoting plant growth (global greening), thereby enhancing agricultural yields, and by neutralizing ocean alkalinity. But the primary concern about CO₂ is its role as a greenhouse gas (GHG) that alters the earth’s energy balance. How the climate will respond to that influence is a complex question that will occupy much of this report.

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2 DIRECT IMPACTS OF CO₂ ON THE ENVIRONMENT

Chapter summary:

CO₂ enhances photosynthesis and improves plant water use efficiency, thereby promoting plant growth. Global greening due to increased CO₂ levels in the atmosphere is well-established on all continents.

CO₂ absorption in sea water makes the oceans less alkaline. The recent decline in pH is within the range of natural variability on millennial time scales. Most ocean life evolved when the oceans were mildly acidic. Decreasing pH might adversely affect corals, although the Australian Great Barrier Reef has shown considerable growth in recent years.

2.1 CO₂ and global greening

The growing CO₂ concentration in the atmosphere has the important positive effect of promoting plant growth by enhancing photosynthesis and improving water use efficiency. That is evident in the “global greening” phenomenon discussed below, as well as in the improving agricultural yields discussed in Chapter 10.

2.1.1 Measurement of global greening

“Greening” refers to an increase in the fraction of the Earth’s surface covered by plants. It can be quantified by the “Leaf Area Index” (LAI) measured by satellite. Many studies over the past decade have confirmed a global greening pattern (increase in LAI) attributable to rising CO₂ levels. Zhu *et al.* (2016) was one of the first studies to report that global greening was detectable using satellite sensors. From 1982 to 2011 they detected greening over 25-50 percent of the Earth versus “browning” over only four percent and attributed 70 percent of the greening to rising CO₂ levels (see Figure 2.1).

Zheng *et al.* (2017) confirmed the pattern, noting that greening over thirty years had added 8 percent to global leaf area and that greening was mitigating warming. Greening has been observed globally but Chen *et al.* (2019) noted that China and India were exploiting it through land management changes. Thus, while China accounts for only 6.6 percent of global vegetated area it accounts for 25 percent of global net increase in LAI. Piao *et al.* (2020) noted that greening was even observable in the Arctic.

While plant models predict increased photosynthesis in response to rising CO₂, Haverd *et al.* (2020) reported a CO₂ fertilization rate nearly double model predictions. That is, CO₂ fertilization had driven an increase in observed global photosynthesis by 30 percent since 1900, versus 17 percent predicted by plant models. This indicates that global models of the socioeconomic impacts of rising CO₂ have likely understated the benefits to crops and agriculture. The connection between CO₂ fertilization and agriculture will be discussed in Chapter 9.

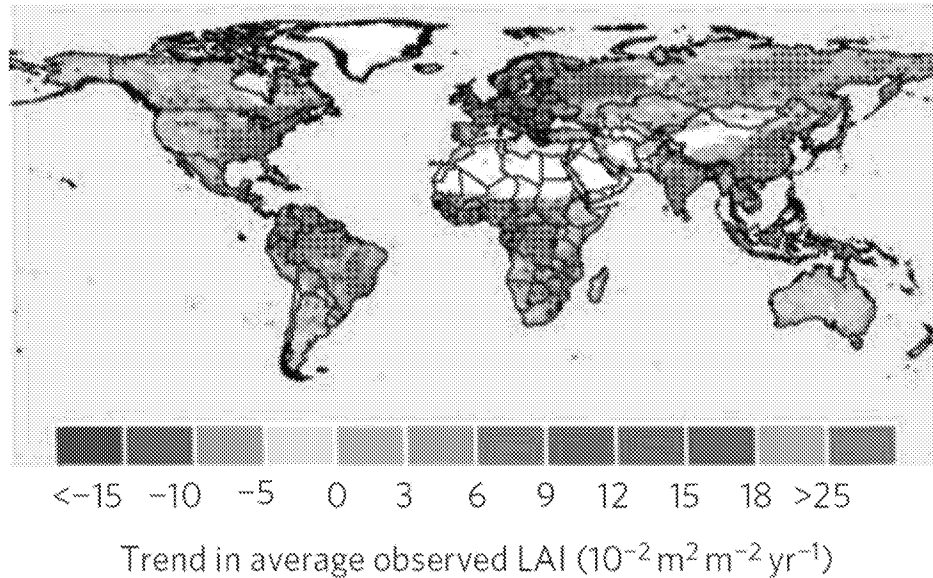


Figure 2.1: Trends in average Leaf Area Index (LAI). Source: Zhu *et al.* 2016 Figure 3.

Finally, Chen *et al.* (2024) surveyed claims that greening had slowed since 2000 due to drought stress. They found instead that faulty satellite sensors were the cause and that the corrected data showed accelerated greening over 55 percent of the globe while browning was accelerating over only 7 percent. In other words, the greening trend continued with no evidence of slowdown, and CO_2 fertilization remained the dominant driver

2.1.2 Photosynthesis and CO_2 levels

Plants build biomass through *photosynthesis*, a process that converts carbon dioxide, water, and light into sugar. The plant enzyme responsible for photosynthesis is Ribulose-1,5-bisphosphate-carboxylase/oxygenase or “Rubisco”. Photosynthesis is initiated when CO_2 is available at the surface of the Rubisco enzyme where it is converted to a molecule with 3 carbon atoms and thereafter incorporated into plant mass. This is referred to as the “C3” process.

Rubisco is estimated to have evolved about 3 billion years ago. Over geological time the Earth’s atmospheric CO_2 levels were usually many times higher than they are today, as shown in Figure 2.2.

About 400 million years ago CO_2 levels were an estimated 2,000-4,000 ppm and were at or above 1,000 ppm for much of the interval from 200 to 50 million years ago (Berner 2006, Judd *et al.* 2024). Over the past 35 million years the level of atmospheric CO_2 has been steadily declining, falling to as low as 170 ppm during glaciations (Gerhart and Ward 2010).

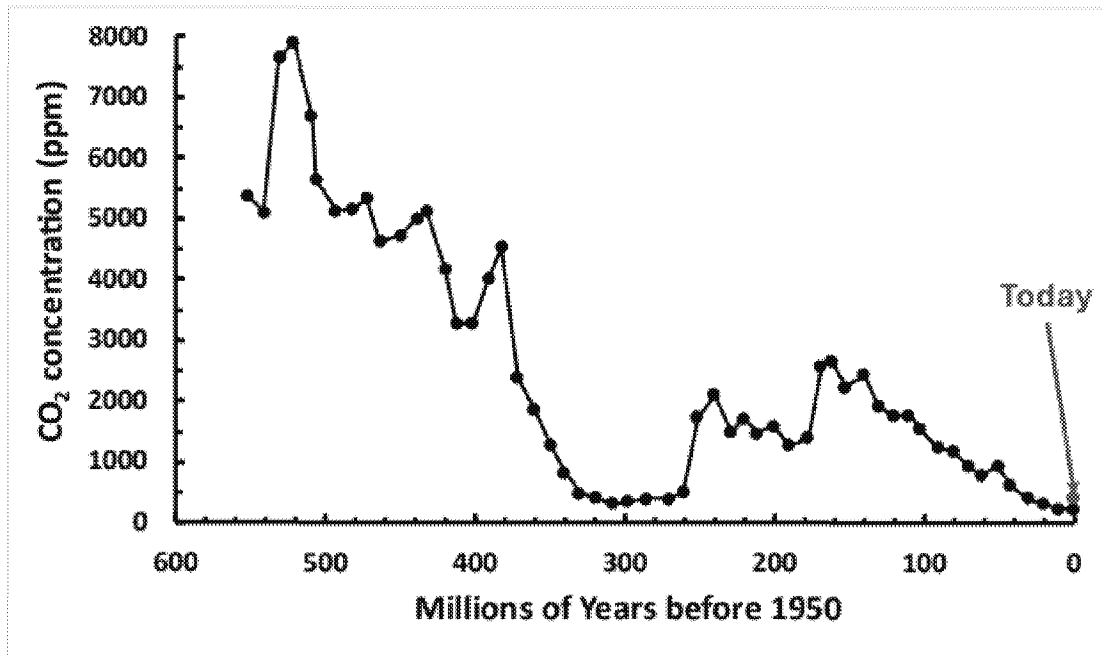


Figure 2.2. Ambient CO₂ levels over the past 550 million years. Today's concentration of 430 ppm is also shown. Source: (Berner 2006)

In response to low-CO₂ conditions some plants evolved another photosynthetic pathway called C₄, in which CO₂ is concentrated in the vicinity of Rubisco, allowing for the C₃ process to function more efficiently. For agricultural purposes the plant categories are:

- C₃: rice, wheat, soybeans and most other crops
- C₄: maize (corn), sugar cane, millet, sorghum

Had atmospheric CO₂ levels continued declining, plant growth would have declined and eventually ceased. Below 180 ppm, the growth rates of many C₃ species are reduced 40-60 percent relative to 350 ppm (Gerhart and Ward 2010) and growth has stopped altogether under experimental conditions of 60—140 ppm CO₂. Some C₄ plants are still able to grow at levels even as low as 10 ppm, albeit very slowly (Gerhart and Ward 2010).

Current CO₂ levels are about 430 ppm, up from 280 ppm in the early 1800s. The positive response of plants to extra CO₂ is illustrated in Figure 2.3, reproduced from Gerhart and Ward (2010). It shows the growth effect of CO₂ on Velvetleaf (*Abutilon theophrasti*) seedlings over 14 days under controlled conditions where only the CO₂ exposure is varied. The gains induced by increasing CO₂ from 150 ppm to 350 ppm continue under a further doubling to 700 ppm.

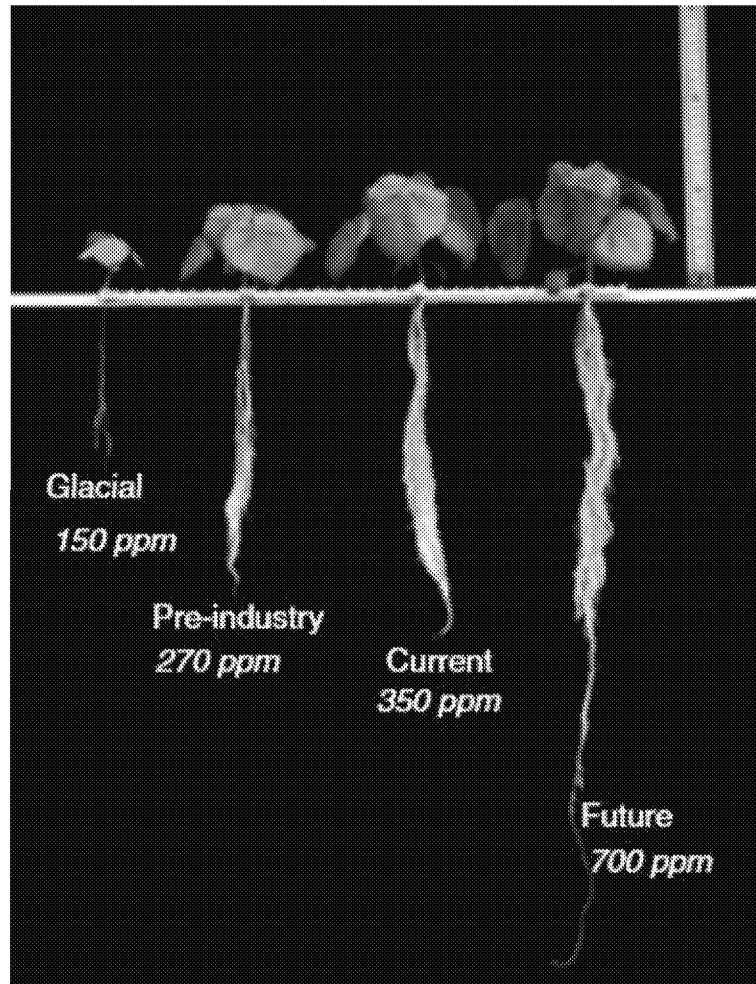


Figure 2.3: growth of *Abutilon theophrasti* after 14 days under identical conditions but for the indicated variations in CO₂ levels. Source: Gerhart and Ward (2010). Note “Current” corresponds to 1988 in image.

Over the past 60+ years there have been thousands of studies on the response of plants to rising CO₂ levels. The overwhelming theme is that plants, especially C3 plants, benefit from extra CO₂. There are two mechanisms by which CO₂ confers a growth benefit:

- Enhanced photosynthesis via the metabolic pathways described above.
- Increased water use efficiency. This arises because plants draw in CO₂ by opening the stomata (pores) on the leaf surface. When CO₂ is scarce the stomata must be kept wide open for long periods, allowing water to evaporate. Under enriched CO₂ conditions the stomata remain closed for longer periods, thus helping the plant retain water longer, and so increasing water use efficiency.

Specific effects of climate change on U.S. agriculture will be reviewed in Chapter 9.

2.1.3 Rising CO₂ and crop water use efficiency

Derying *et al.* (2016) surveyed evidence on crop water productivity (CWP), the yield per unit of water used, drawing attention to the potential for CO₂ both to enhance photosynthesis and to reduce leaf-level transpiration (water loss during leaf respiration). They surveyed all available FACE data (Free Air CO₂ Enrichment—see Chapter 9) on crop yield changes for maize (corn), wheat, rice, and soybean and combined

it with crop model data simulating yield responses as of 2080 under the extreme RCP8.5 emissions scenario in four growing regions (Tropics, Arid, Temperate and Cold) each of which were split into rainfed and irrigated sub-regions. They reported that models without CO₂ fertilization predicted CWP losses in every region, but those were more than offset by CO₂ fertilization so that all regions showed a net CWP gain. Deryng *et al.* (2016) also reported that negative impacts of warming on wheat and soybean yields were fully offset by CWP gains and mitigated by up to 90 percent for rice and 60 percent for maize.

Similarly, Cheng *et al.* (2017) noted that increased global plant growth from 1982 to 2011 due to rising CO₂ uptake was accompanied by such large gains in CWP that global water use by plants had not increased, despite the extra biomass.

Deryng *et al.* (2016) assumed that climate change would “exacerbate water scarcity”. Yet while models do predict that drylands will expand under climate warming, current data show the opposite: greening is happening even in arid areas. Zhang *et al.* (2024) report that due to increased CO₂ levels “increasing aridity in drylands won’t lead to a general loss of vegetation productivity”; at most only 4 percent of currently arid areas will see increased desertification.

2.1.4 CO₂ fertilization benefits in IPCC Reports

The IPCC has only minimally discussed global greening and CO₂ fertilization of agricultural crops. The topic is briefly acknowledged in a few places in the body of the IPCC 6th and earlier Assessment Reports but is omitted in all Summary documents. Section 2.3.4.3.3 of the AR6 Working Group I report, entitled “global greening and browning,” points out that the IPCC Special Report on Climate Change and Land had concluded with *high confidence* that greening had increased globally over the past 2-3 decades. It then discusses that there are variations in the greening trend among data sets, concluding that while they have *high confidence* greening has occurred, they have *low confidence* in the magnitude of the trend. There are also brief mentions of CO₂ fertilization effects and improvements in water use efficiency in a few other chapters in the AR6 Working Groups I and II Reports.

Overall, however, the Policymaker Summaries, Technical Summaries, and Synthesis Reports do not discuss the topic.

2.2 The Alkaline Oceans

2.2.1 Changing pH

A neutral aqueous solution has a pH of 7.0, while one with pH greater than 7.0 is alkaline (also termed basic) and with pH less than 7.0 is acidic. The modern-day global average pH of sea water is estimated to be 8.04 (Copernicus Marine Service 2025), down from an estimated value of 8.2 in pre-Industrial times (Gattuso and Hansson, 2011). As CO₂ concentrations in the atmosphere increased, the oceans absorbed more, which decreases their pH. Depending upon the oceans’ buffering capacity, they are expected to become somewhat less alkaline over time, consistent with the observed decrease in pH.

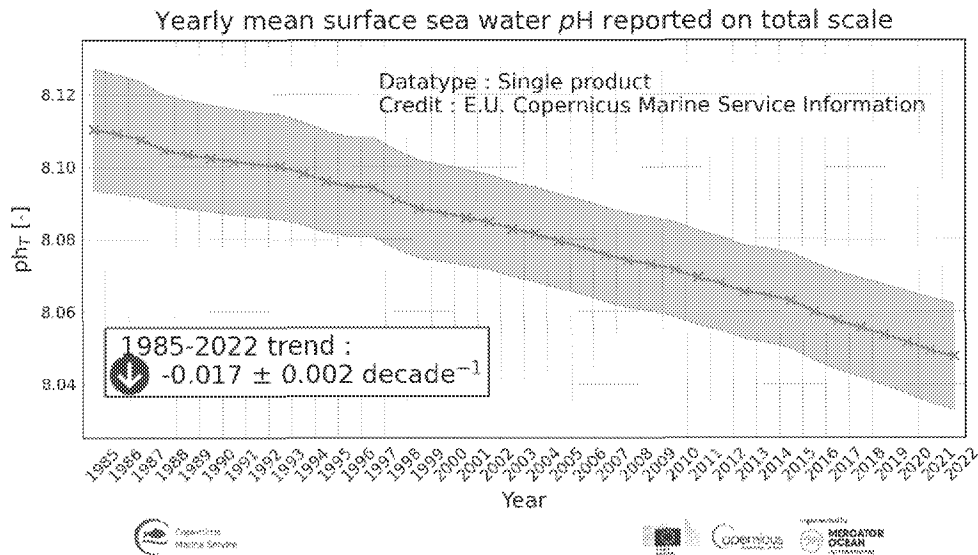


Figure 2.4: Ocean pH 1985 – 2022. Source: Copernicus Marine Service 2025

While this process is often called “ocean acidification”, it is a misnomer because the oceans are not expected to become acidic; “ocean neutralization” would be more accurate. Even if the water were to turn acidic, it is believed that life in the oceans evolved when the oceans were mildly acidic with pH 6.5 to 7.0 (Krissansen-Totton *et al.*, 2018). On the time scale of thousands of years, boron isotope proxy measurements show that ocean pH was around 7.4 or 7.5 during the last glaciation (up to about 20,000 years ago) increasing to present-day values as the world warmed during deglaciation (Rae *et al.*, 2018). Thus, ocean biota appear to be resilient to natural long-term changes in ocean pH since marine organisms were exposed to wide ranges in pH.

2.2.2 Coral reef changes

There are concerns that a decreasing pH of sea water will reduce the calcification rate of coral reefs. But coral reefs already endure large swings in pH, partly due to daily photosynthetic activity in the reef; measured pH values range from 9.4 during the day to 7.5 at night (Revelle and Fairbridge, 1957). De’ath *et al.* (2009) reported that a portion of Australia’s Great Barrier Reef (GBR, the world’s largest coral reef ecosystem) had experienced a 14 percent decline in calcification since 1990. This was tentatively attributed to increasing water temperature and decreasing pH. But Ridd *et al.* (2013) showed that report to have resulted from a biased data analysis that, when corrected, showed no change in calcification rates. Nevertheless, the alarm produced by the original paper has persisted as evidenced by the large number of published citations (541) to the original study compared to only 11 citations to the correction (as of 30 April 2025).

The most recent annual summary of GBR conditions from the Australian Institute of Marine Science indicates that coral production has rebounded strongly (AIMS, 2023). Figure 2.5 shows the results of the AIMS surveys of hard coral cover, expressed as a percentage of the reef area. Much of the decline in the GBR before 2011 turned out to be due to intense tropical cyclone activity (Beeden *et al.*, 2015) as well as a string of marine heatwaves, agricultural runoff and invasive species (Woods Hole, 2023). Given the reported declines in GBR calcification between 1990 and 2009 and the continued increase in atmospheric CO₂ levels the rebound has surprised some observers.

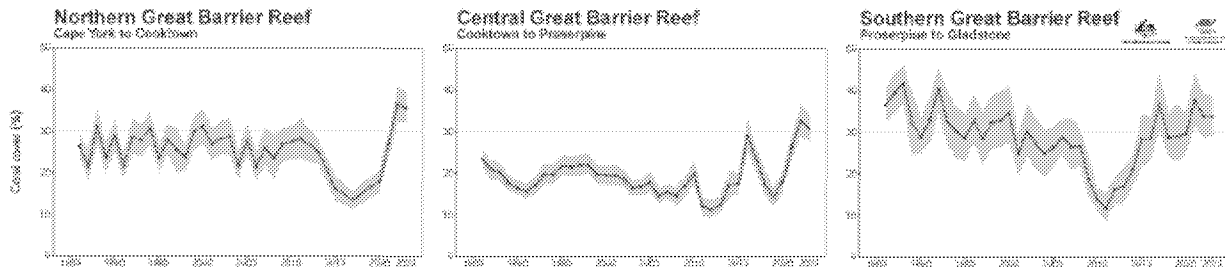


Figure 2.5 Hard coral cover of three regions of the Great Barrier Reef 1985 to 2023. Source: AIMS 2023.

It is being increasingly recognized that publication bias (alarming ocean acidification results preferred by high-impact research publications) exaggerates the reported impacts of declining ocean pH. An ICES Journal of Marine Science Special Issue addressed this problem with an article entitled, *Towards a Broader Perspective on Ocean Acidification Research*. In the Introduction to that Special Issue, H. I. Browman stated, “As is true across all of science, studies that report no effect of ocean acidification are typically more difficult to publish.” (Browman, 2016).

Similarly, a meta-analysis (Clements *et al.*, 2021) of the negative effects of ocean acidification on reef fish behavior found what they called a “decline effect”: initially dramatic conclusions published in prominent journals showing apparently large impacts of acidification tended to be followed up by subsequent studies on larger sample sizes yielding much smaller and typically non-existent effects. They call for their colleagues to improve research practices to counter the “decline effect”:

[The] vast majority of studies with large effect sizes in this field tend to be characterized by low sample sizes, yet are published in high-impact journals and have a disproportionate influence on the field in terms of citations. We contend that ocean acidification has a negligible direct impact on fish behavior, and we advocate for improved approaches to minimize the potential for a decline effect in future avenues of research (Clements *et al.*, 2021).

In summary, ocean life is complex and much of it evolved when the oceans were acidic relative to the present. The ancestors of modern coral first appeared about 245 million ago; as can be seen from Figure 2.2, CO₂ levels for more than 200 million years afterward were many times higher than they are today. Much of the public discussion of the effects of ocean “acidification” on marine biota has been one-sided and exaggerated. There is a known tendency for studies reporting dramatic effects on things like coral cover and fish behavior to get published in high-impact journals and receive widespread press coverage, then subsequent research rebutting those findings receives little attention.

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3 HUMAN INFLUENCES ON THE CLIMATE

Chapter Summary:

The global climate is naturally variable on all time scales. Anthropogenic CO₂ emissions add to that variability by changing the total radiative energy balance in the atmosphere.

The IPCC has downplayed the role of the sun in climate change but there are plausible solar irradiance reconstructions that imply it contributed to recent warming.

Climate projections are based on IPCC emission scenarios that have tended to exceed observed trends. Most academic climate impact studies in recent years are based upon the extreme RCP 8.5 scenario that is now considered implausible; its use as a business-as-usual scenario has been misleading.

Carbon cycle models connect annual emissions to growth in the atmospheric CO₂ stock. While models disagree over the rate of land and ocean CO₂ uptake, all agree that it has been increasing since 1959.

3.1 Components of radiative forcing and their history

3.1.1 Historical radiative forcing

A changing climate has been the norm throughout the Earth's 4.6-billion-year history. The Earth's temperature and weather patterns change naturally over timescales ranging from decades to millions of years. Natural variations in the surface climate originate in two ways. Internal climate fluctuations associated with circulations in the atmosphere and ocean exchange energy, water, and carbon between the atmosphere, oceans, land, and ice. External influences on the climate system include variations in the energy received from the sun and the effects of volcanic eruptions. Human activities influence climate through changing land use and land cover. Humans are also changing the composition of the atmosphere by emissions of CO₂ and other greenhouse gases and by altering the concentration of aerosol particles in the atmosphere.

The earth is warmed by the sunlight it absorbs and is cooled by the heat it radiates to space. Averaged over the Earth's surface, each of these processes involve power flows of about 240 Watts per square meter (W/m²). When they are in balance, there are no net external causes of warming or cooling. Both human and natural influences on the climate alter this balance and so cause the climate to change.

Influences on the Earth's energy balance at the top of the atmosphere are quantified by "radiative forcing", the extent to which they disturb the warming/cooling balance; a positive forcing warms while a negative forcing cools. The IPCC's estimated history of major components of radiative forcing since 1750 is shown in the following two figures from its AR6.

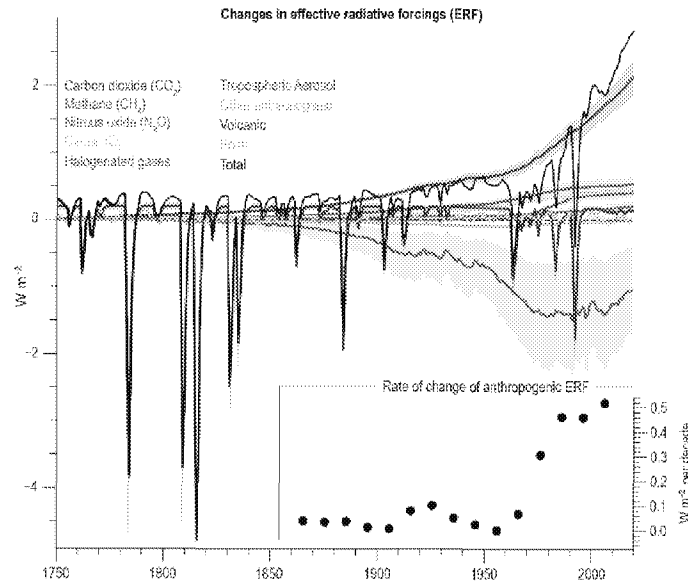


Figure 3.1.1: IPCC estimates of radiative forcing components over time. Shading indicates uncertainty ranges. Source: AR6 WGI Ch2 Fig. 10

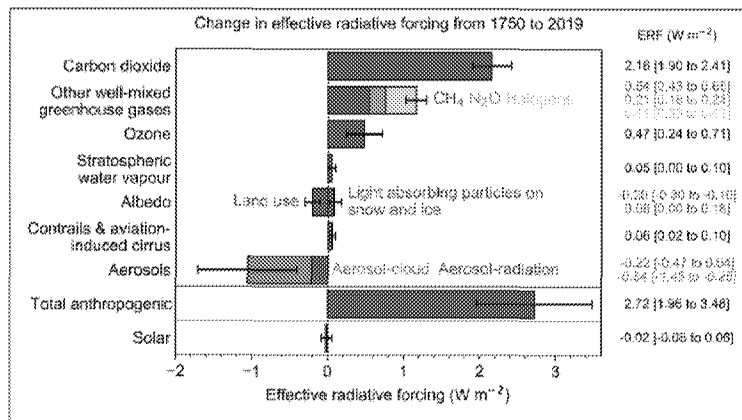


Figure 3.1.2 IPCC estimates of radiative forcing component changes from 1750 to 2019. Source: AR6 WGI Ch 7 Fig. 7-6.

These graphs show that the total radiative forcing is comprised of both natural and anthropogenic components. Carbon dioxide is the largest human influence on the climate and the one most relevant to the influence of fossil fuel use. It exerts a warming influence by decreasing the cooling power of the atmosphere. Emissions of CO_2 are accumulating in the atmosphere, as described in the following section, so that the warming influence is growing. Other greenhouse gases (methane, nitrous oxide, halogens, and ozone) act similarly, currently adding another 75 percent to CO_2 's warming. Aerosols exert an overall cooling effect, although with large uncertainties in the way they catalyze the formation of reflective clouds. As a result, understanding the causes of recent warming requires not just identifying the warming effects of CO_2 , but also the more uncertain cooling effects of aerosols.

The IPCC assesses the change in the radiative forcing by the sun to be negligible, based on their preference for data reconstructions that imply minimal solar change since preindustrial times. But Connolly *et al.* (2021) reviewed sixteen different Total Solar Irradiance (TSI) reconstructions in the literature

covering the years 1600-2000; the reconstructions vary from almost no change in TSI to a relatively large upward trend. Those authors note that the variation in TSI reconstructions combined with variations in surface temperature reconstructions allows for inferences consistent with either no or most 20th century warming being attributable to the sun.

A particularly thorny issue is the gap in TSI data between 1989 and 1991 due to a delay in the launch of a monitor following the Space Shuttle Challenger disaster on January 28 1986. This delay prevented a replacement satellite from being launched in time to overlap with, and its readings to be intercalibrated with, the prior system (Zacharias 2014, Scafetta *et al.* 2019). This is called the ACRIM (Active Cavity Radiometer Irradiance Monitor) gap problem. The question of whether there is an upward trend in TSI over 1978 to 2018 hangs on how the ACRIM data gap is filled. Connolly *et al.* (2021) found that the IPCC's consensus statements on solar forcing were formulated prematurely through the suppression of dissenting scientific opinions.

Another natural radiative forcing component is volcanic aerosols, which exert an episodic cooling influence. Box 4.1 in the IPCC AR6 Report addresses the climate impact of volcanic eruptions, noting three explosive volcanic eruptions that occurred in the first half of the 19th century. This included the 1815 Tambora eruption that resulted in the ‘year without summer’, with multiple harvest failures across the Northern Hemisphere. There is uncertainty about the sign of the relatively small forcing due to the submarine volcano Hunga Tonga which erupted in 2022 (Jenkins *et al.* 2023, Schoeberl *et al.* 2024).

Figure 3.1.1 shows that the anthropogenic forcing component was negligible before about 1900 and has increased steadily since, rising to almost 3 W/m² today. However, this is still only about 1 percent of the unperturbed radiation flows, making it a challenge to isolate the effects of anthropogenic forcing; state-of-the-art satellite estimates of global radiative energy flows are only accurate to a few W/m².

Natural sources of global energy imbalance other than volcanoes and total solar irradiance (TSI) are not included in these graphs because they remain largely unknown.

3.1.2 Change in atmospheric CO₂ since 1958

Carbon dioxide's warming influence depends on how much “extra” CO₂ accumulates in the atmosphere- *i.e.*, its concentration above the preindustrial value of 280 ppm. The CO₂ level as recorded at the Mauna Loa observatory in Hawaii, generally used as the representative global average concentration, is available online at [[HYPERLINK "https://gml.noaa.gov/ccgg/trends/index.html"](https://gml.noaa.gov/ccgg/trends/index.html)]. The concentration was about 316 ppm at the start of the record in 1959 and is now about 430 ppm, a 36 percent increase. At the end of the last glaciation CO₂ levels had fallen to about 180 ppm. As discussed in Chapter 2, C3 plants begin dying at CO₂ levels below about 140 ppm and C4 plants at levels below 100 ppm, so if CO₂ levels had continued falling plant life would have been imperiled.

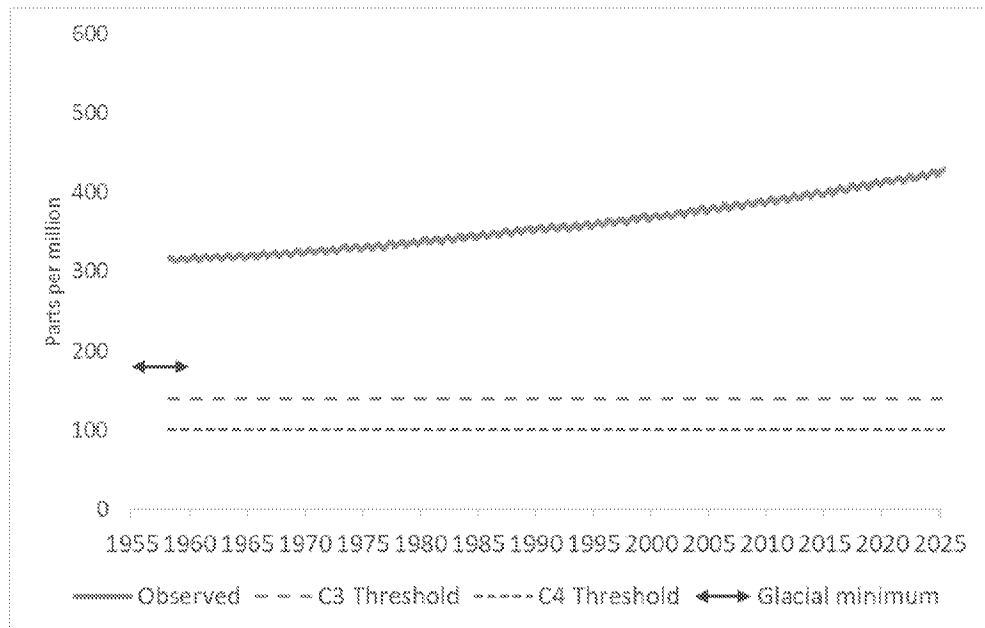


Fig. 3.1.3. Yearly average atmospheric CO₂ concentrations (1959-2025) in ppm measured at Mauna Loa (blue). C3 Threshold: Level below which C3 plants begin dying (140 ppm, see Chapter 2). C4 Threshold: Level below which C4 plants begin dying (100 ppm, see Chapter 2). Glacial minimum: Minimum level during recent glaciations (purple arrow). CO₂ data source: [HYPERLINK "<https://gml.noaa.gov/ccgg/trends/index.html>"]

The annual increase in concentration is only about half of the CO₂ emitted because land and ocean processes currently absorb “excess” CO₂ at a rate approximately 50 percent of the human emissions. Future concentrations, and hence future human influences on the climate, therefore depend upon two components: (1) future rates of global human CO₂ emissions, and (2) how fast the land and ocean remove extra CO₂ from the atmosphere. We discuss each of these in turn.

3.2 Future emission scenarios and the carbon cycle

3.2.1 Emission scenarios

Assessing the dangers of future GHG emissions requires assumptions about what those emissions will be. Future emissions, and hence human influences on the climate, will depend upon future demographics, economic activity, regulation, and energy and agricultural technologies. Various assumptions about each of those lead to projections of greenhouse gas emissions and concentrations, aerosol concentrations, and changes in land use, which ultimately can be combined into assumptions about anthropogenic radiative forcing.

The great uncertainties about these many factors make it impossible to precisely predict future emissions. Instead, the IPCC has created various sets of scenarios meant to span a plausible range of possibilities for population, economy, and technologies. Recent versions of the scenarios are labeled by a number indicating the anthropogenic radiative forcing expected in 2100 under that scenario. Thus, a scenario labeled with a “6” corresponds to 6 W/m² of human-induced radiative forcing (warming) at the end of the century. (Recall current anthropogenic radiative forcing is about 2.7 W/m².)

Although the IPCC does not claim its emission scenarios are forecasts, they are often treated as such. Comparisons of past scenario groups against observations show that IPCC emission projections tend to overstate actual subsequent emissions. For the IPCC Third and Fourth Assessment Reports a set of emission projections from the *Special Report on Emission Scenarios* was used; these were referred to as the SRES scenarios. McKittrick *et al.* (2012) showed that, when converted to per capita values, the SRES scenario emissions distribution was skewed upwards and incompatible with observed trends. McKittrick and Wood (2013) further showed that global energy market dynamics constrained global per capita CO₂ emissions to an extent not reflected in IPCC scenarios. The bias of the SRES scenarios was confirmed by the later analysis of Hausfather *et al.* (2019) who showed that observed atmospheric CO₂ concentrations tracked the low end of the SRES range and also of subsequent IPCC scenario ranges (Figure 3.2.1).

For AR5 the IPCC developed a new set of scenarios called the Representative Concentration Pathways (RCPs). These were identified by a number representing the increase in forcing and were thus called RCP2.6, RCP4.5, RCP6.0 and RCP8.5. RCP2.6 (implying an anthropogenic radiative forcing in 2100 of 2.6 W/m²) describes a GHG concentration pathway leading to warming well below 2°C. At the other end of the scale RCP8.5 is an extreme outcome implying nearly 5°C warming from 1900 to 2100.

RCP8.5 came to be referred to as the no-policy baseline, or “business-as-usual” scenario in both the academic literature and popular media. It was therefore used to generate the reference outcome supposedly representing the 21st century world in the absence of increasingly stringent emission reduction policies. But RCP8.5 was intended as a low-probability high emissions scenario and its use as a business-as-usual baseline has been criticized as grossly misleading.¹ Hausfather and Peters (2020) writing in a commentary in *Nature*, pointed out that RCP8.5 was developed as an extreme worst-case, and its misuse as a “business as usual” baseline has resulted in a large number of misleading studies and media reporting.

The implausibility of the RCP8.5 scenario was examined by Burgess *et al.* (2021). They noted that RCP8.5 has already diverged from observed trends in energy use and the near future trends diverge sharply from those of the International Energy Agency (IEA), which provides market-based projections of energy use for the coming decades. The IEA trends, which are based on plausible extrapolations of current economic and demographic trends, run below the entire envelope of RCP projections, implying not only that RCP8.5 is dubious, but that even lower emission scenarios are unlikely.

Widespread use of RCP8.5 has created a bias towards alarm in the climate impacts literature. The extent of this problem was confirmed in a literature analysis by Pielke Jr. and Ritchie (2020). They found that some 16,800 scientific papers published between 2010 and 2020 used the RCP8.5 scenario, with about 4,500 of the articles linking RCP8.5 to the concept of “business-as-usual”. Their analysis showed how RCP8.5 was misused not only by individual researchers, but also by influential science agencies like the IPCC and the U.S. National Climate Assessment (USNCA), which has directly led to misleading coverage in prominent media outlets.

¹ This extreme scenario is useful for modelers, since a large forcing generates a large response (warming) making it easier to assess a model’s sensitivity. But that’s very different than claiming it is a plausible future outcome.

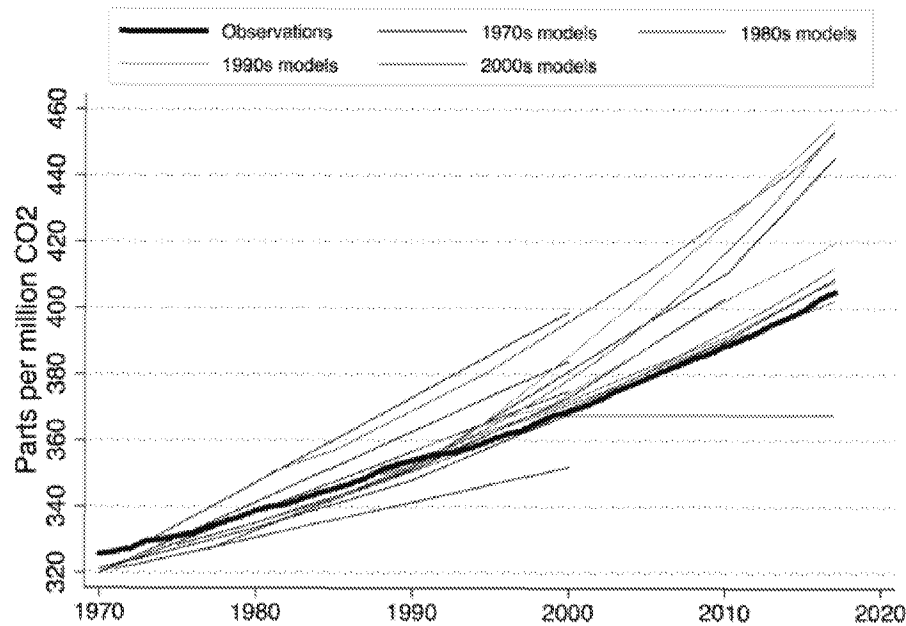


Figure 3.2.1 Comparison of observed atmospheric CO₂ levels (black line) with scenario projections (colored lines, mainly IPCC) since the 1970s. Source: Hausfather *et al.* (2019) Figure S4.

Pielke and Ritchie (2020) reported new studies using RCP8.5 were published at a rate of about 20 per day, with about two per day specifically linking RCP8.5 and “business as usual.” They conclude that the climate research community has spent a decade “committing scientific resources to science fiction” and that “The scientific literature has become imbalanced in an apocalyptic direction.”

The IPCC developed a new set of scenarios for AR6, the “Shared Socioeconomic Pathway” (SSP) scenarios, which have continued the bias shown in the RCP and SRES scenarios. Figure 3.2.2 shows total global observed CO₂ emissions compiled by the International Energy Agency (IEA) merged with the emission projection from the IEA taking account of energy use projections and current policies. The other lines show the range of IPCC SSP scenarios (SSP1-SSP5). As of 2023, global CO₂ emissions have been tracking well below SSP7.0 and are even below SSP2-4.5.

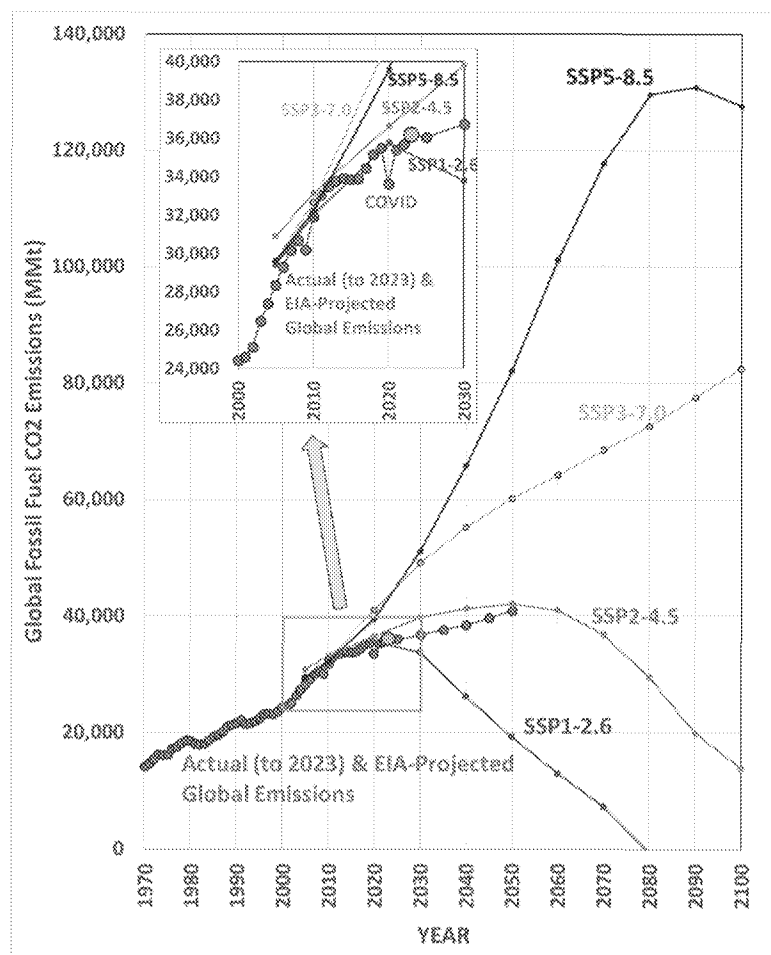


Figure 3.2.2. Observed and projected CO₂ emissions. Source: IPCC (SSP scenarios) and Energy Information Administration (EIA). Green: observed historical emissions and EIA projections. Other lines: SSP1-5. Data source: Friedlingstein *et al.* (2024).

3.2.2 The carbon cycle relating emissions and concentrations

Carbon dioxide emissions from fossil fuel burning (and to a lesser extent deforestation and cement production) have led to steadily increasing CO₂ concentrations in the atmosphere, as shown in Fig. 3.1.3. The relation between emissions and concentration is determined by the global carbon cycle of land and ocean processes that exchange carbon with the atmosphere. Our understanding of these processes was reviewed by Crisp *et al.* (2021).

There are about 850 Gt of carbon (GtC) in the Earth's atmosphere², almost all of it in the form of CO₂. Each year, biological processes (plant growth and decay) and physical processes (ocean absorption and outgassing) exchange about 200 GtC of that carbon with the Earth's surface (roughly 80 GtC with the land and 120 GtC with the oceans). Before human activities became significant, removals from the atmosphere were roughly in balance with additions. But burning fossil fuels (coal, oil, and gas) removes carbon from the ground and adds it to the annual exchange with the atmosphere. That addition (together with a much

² Because CO₂ is chemically transformed through the course of the carbon cycle, it is more convenient to track atoms of carbon rather than molecules of CO₂. One gigatonne of carbon (GtC) is equivalent to about 3.7 Gt of CO₂.

smaller contribution from cement manufacturing) amounted to 10.3 GtC in 2023, or only about 5 percent of the annual exchange with the atmosphere.

The carbon cycle accommodates about 50 percent of humanity’s small annual injection of carbon into the air by naturally sequestering it through plant growth and oceanic uptake, while the remainder accumulates in the atmosphere (Ciais *et al.*, 2013). For that reason, the annual increase in atmospheric CO₂ concentration averages only about half of that naively expected from human emissions.

To project future CO₂ concentrations in the atmosphere, and hence future human influences on the climate, it is important to know how the carbon cycle might change in the future. The historical near constancy of that 50 percent fraction means that the more CO₂ humanity has produced, the faster nature removed it from the atmosphere. That 50 percent fraction changes somewhat from year to year due to natural carbon cycle imbalances from El Nino, La Nina, and varying weather patterns. There was also a substantial additional reduction in atmospheric CO₂ after the 1991 eruption of Mt. Pinatubo, a curious result that has yet to be explained (Angert *et al.*, 2004).

The main processes that remove excess CO₂ from the atmosphere are increased growth of land vegetation (especially at high latitudes), some increase in the sequestering of carbon in soils, and uptake of CO₂ by the ocean due to the increasing partial pressure of atmospheric CO₂ over that of CO₂ dissolved in the ocean. All twenty land carbon cycle models tracked by the Global Carbon Project (Friedlingstein *et al.*, 2024) show land processes have been removing excess CO₂ at an increasing rate since 1959. This is consistent with a “global greening” phenomenon (Chapter 2.1) observed by satellites since monitoring of global greenness began in 1982.

While land vegetation has been responding positively to more atmospheric CO₂, uptake of extra CO₂ by ocean biological processes remains too uncertain to be measured reliably. Our current understanding of these and many more carbon cycle processes was reviewed by Crisp *et al.* (2021).

CO₂ uptake by land processes

The uptake of extra CO₂ from the atmosphere by land surface processes (as also inferred from global greening) has been modeled with 20 different dynamic global vegetation models, the outputs of which are updated every year by the Global Carbon Project (Friedlingstein, 2024). As seen in Fig. 3.2.3, all of those models agree that vegetation and soils have been sequestering carbon from the atmosphere. But we also see that the long-term trends over 1959 to 2023 (65 years) vary widely between models, by nearly a factor of 7. This demonstrates that there remains considerable uncertainty in how fast land processes are removing CO₂ from the atmosphere, which in turn creates uncertainty in future atmospheric CO₂ concentrations, which then produce uncertainty in climate model simulations of future climate change.

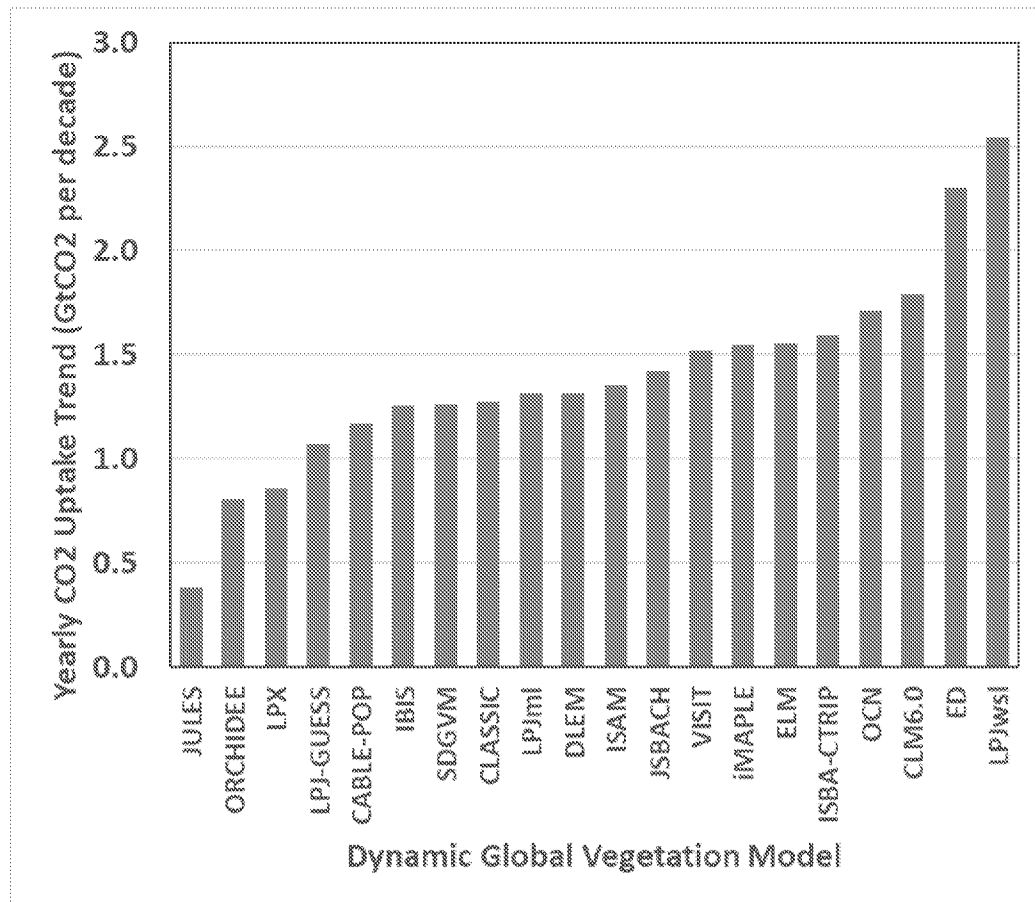


Figure 3.2.3 Trends of annual CO₂ uptake (GtCO₂ per year per decade) by land processes during 1959-2023 simulated by 20 different dynamic global vegetation models periodically reported by the Global Carbon Project (Friedlingstein, 2024).

CO₂ uptake by ocean processes

The uptake of extra CO₂ from the atmosphere by ocean processes has been modeled with 10 different ocean biogeochemistry models, the outputs of which are updated every year by the Global Carbon Project (Friedlingstein, 2024). Like the results from the land models, all of the ocean models agree that the global oceans have been sequestering carbon from the atmosphere at an increasing rate during 1959-2023 (Fig. 3.2.4). Unlike the land models, however, the ocean models show much better agreement with each other, with the model producing the fastest increasing CO₂ uptake being only 65 percent faster than the model with the most slowly increasing CO₂ uptake. In spite of the relative agreement among models, Friedlingstein *et al.* (2022) notes that there is substantial discrepancy between the different methods on the strength of the ocean sink over the last decade, particularly in the Southern Ocean.

Note that the average trend in CO₂ uptake across all land models in Fig. 3.2.3 is 25 percent larger than the average trend in ocean uptake. This suggests land processes are increasing in their ability to remove CO₂ faster than ocean processes are increasing their CO₂ sequestration.

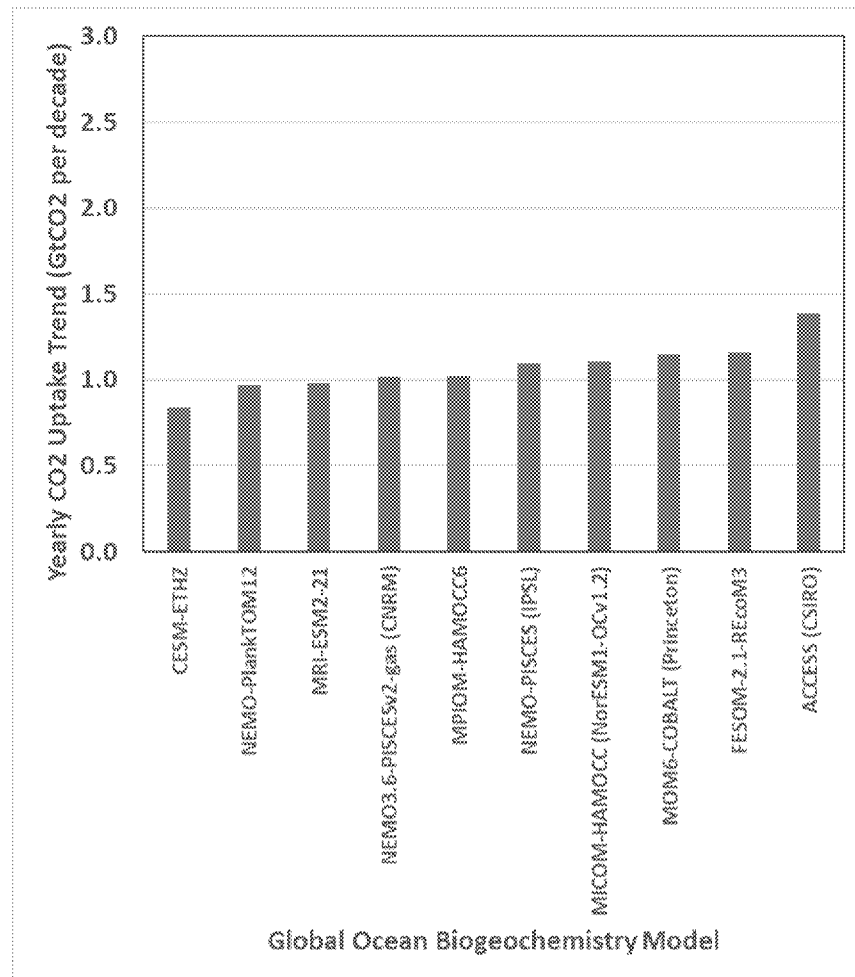


Figure 3.2.4 Trends of annual CO₂ uptake (GtCO₂ per year per decade) by ocean processes during 1959-2023 simulated by 10 different ocean biogeochemistry models periodically reported by the Global Carbon Project (Friedlingstein, 2024).

3.3 Urbanization influence on temperature trends

Historical temperature data over land has been collected mainly where people live. This raises the problem of how to filter out non-climatic warming signals due to Urban Heat Islands (UHI) and other changes to the land surface. If these are not removed the data might over-attribute observed warming to greenhouse gases. The IPCC acknowledges that raw temperature data are contaminated with UHI effects but claims to have data cleaning procedures that remove them. It is an open question whether those procedures are sufficient.

AR6 downplayed this issue by saying (WGI p. 235) that no recent evidence had emerged to alter the AR5 finding that urbanization causes an upward bias of no more than 10 percent in the global land surface warming trend. AR5 (WGI p. 189) also cited the 10 percent upper bound without citing a source. AR4 (WGI p. 244) cited Jones *et al.* (1990) and Peterson *et al.* (1999) as the basis of the claim. Peterson *et al.* had failed to find any difference in trends between rural and urban samples, although their definition of rural included local populations up to 10,000 persons while the relative influence of urbanization begins well below that (Spencer *et al.*, 2025). Jones *et al.* compared rural/urban warming in three regions: Eastern Australia, Eastern China and Western Soviet Union. Their definition of “rural” included towns of up to

10,000 in the former Soviet Union and up to 100,000 in China. They found relative warming biases greater than 10 percent in these areas but conjectured that the urbanization effect averaged over the areas they did not examine would bring the global land bias to under 10 percent of the observed warming trend.

Several papers appeared in print prior to the IPCC AR4 that argued that the warming effect of UHIs added a relatively large (30-50%) component to observed warming and was not simulated by climate models (de Laat and Maurellis 2006, McKittrick and Michaels 2007). These findings were based on correlations between locations of maximum warming over land with locations of maximum socioeconomic development. AR4 asserted (p. 244) that these correlations were an artefact of natural atmospheric circulations and were in fact statistically insignificant, and on that basis set the findings aside. Their claim was controversial because it was presented with no supporting evidence. McKittrick (2010) and McKittrick and Nierenberg (2010) showed that taking into account various conjectured alternative explanations for the correlations did not affect their significance. AR5 (p. 189) conceded that AR4 had provided “no explicit evidence” for its assessment and further acknowledged, based on these papers, that there was “significant evidence for such contamination of the record” i.e. a warming bias in the land record. However as already noted, elsewhere in the AR5 report they carried forward AR4 claim that it was less than 10 percent of observed warming. Further they provided no caution about using the land record for climate measurement despite conceding the evidence for UHI contamination. Recently Soon *et al.* (2023) estimated an urbanization bias in the Northern Hemisphere land record over 1850-2018 sufficient to increase the trend in the blended record from 0.55°C to 0.89°C per century.

Some studies providing evidence against UHI contamination compared warming rates between rural and urban locations (Jones *et al.* 1990, Peterson *et al.* 1999, Wickham *et al.* 2013). It is not known whether such methods would be capable of detecting UHI bias even when present. The influence of UHI warming is logarithmic in population, in other words it is strongest at low population density then levels off as local urbanization expands (Oke 1973, Spencer *et al.* 2025). Hence failure to find a difference in warming rates between urban and rural stations does not prove the absence of UHI contamination. McKittrick (2013) provided an empirical demonstration in which the rural/urban trends were not significantly different in a data set shown on other grounds to be contaminated with UHI bias.

Parker (2006) examined a sample of urban locations and found no difference in trends between subsets partitioned according to nighttime wind speed, concluding on this basis that urbanization could not be a significant factor. Here again the question is whether such a method would find UHI bias even if present. McKittrick (2013) presented an example in which UHI-contaminated data did not exhibit significant trend differences when stratified according to wind speed.

The challenge in measuring UHI bias is relating local temperature change to a corresponding change in population or urbanization, rather than to a static classification variable such as rural or urban. Spencer *et al.* (2025) used newly available historical population archives to undertake such an analysis and found evidence of significant UHI bias in U.S. summertime temperature data.

In summary, while there is clearly warming in the land record, there is also evidence that it is biased upward by patterns of urbanization and that these biases have not been completely removed by the data processing algorithms used to produce climate data sets.

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PART II: CLIMATE RESPONSE TO CO₂ EMISSIONS

4 CLIMATE SENSITIVITY TO CO₂ FORCING

Chapter Summary

There is growing recognition that climate models are not fit for the purpose of determining the Equilibrium Climate Sensitivity (ECS) of the climate to increasing CO₂. The IPCC has turned to data-driven approaches including historical data and paleoclimate reconstructions, but their reliability is diminished by data inadequacies.

Data-driven ECS estimates tend to be lower than climate model-generated values. The IPCC AR6 upper bound for the likely range of ECS is 4.0°C, lower than the AR5 value of 4.5°C. This lowering of the upper bound seems well justified by paleoclimatic data. The IPCC AR6 lower bound for the likely range of ECS is 2.5°C, substantially higher than the AR5 value of 1.5°C. This raising of the lower bound is less justified; evidence since the AR6 finds the lower bound of the *likely* range to be around 1.8°C.

4.1 Introduction

The climate's response to increasing concentrations of CO₂ is central to the scientific debate on anthropogenic climate change, and so also to the public debate on “climate action.” The simplest measure of that response is the rise in the global average surface temperature, quantified by the Equilibrium Climate Sensitivity (ECS). ECS is defined as the amount of warming expected in response to a doubling of CO₂ from its pre-industrial concentration of 280 ppm, after all climate components have had time to adjust. Some components, like temperatures in the lower atmosphere (troposphere), adjust rapidly, while others such as the deep ocean and cryosphere might take as long as centuries. A related measure, the Transient Climate Response (TCR), better describes the shorter time scales; it is defined as the amount of warming when the CO₂ concentration is doubled by rising one percent annually for 70 years.

The 1979 Charney Report for the U.S. National Academy of Sciences (National Research Council 1979) proposed that the most likely ECS was $3.0 \pm 1.5^\circ\text{C}$. The IPCC repeatedly reaffirmed that range with only minor variations until its most recent AR6. AR5 termed 1.5–4.5°C as the *likely* range (66 percent probability) and stated that ECS is *extremely unlikely* (95 percent probability) to be below 1.0°C and *very unlikely* (90 percent probability) to exceed 6.5°C.

The uncertainty in ECS has remained stubbornly wide, despite many individual studies that claimed to narrow it (Hausfather 2023). Most recently, AR6 narrowed the *likely* range to 2.5–4.0°C and deemed the *very likely* range to be 2.0–5.0°C. This narrowing on the low end is disputed, as will be discussed below.

Uncertainties in ECS are highly consequential for policy making. As will be discussed in Chapter 11, economic models use ECS values to project the costs of CO₂ emissions. The traditional value (3.0 °C) has typically yielded modest global social costs of CO₂ emissions, sufficient to justify some policy actions, but mostly deferred to later in this century. If ECS is very high (above 4.5°C) immediate aggressive emission controls become more imperative, whereas no CO₂ emission controls are economically justifiable for ECS below 2.0°C (Dayaratna *et al.* 2016, 2020). Obtaining a precise estimate is impossible, so policy making needs to account for the uncertainty.

By itself, the equilibrium warming effect of a doubling of atmospheric CO₂ is slightly more than 1°C (Soden and Held 2006). Larger values of ECS arise from positive feedbacks that amplify the CO₂ warming. Water vapor feedback is positive: a warmer atmosphere might have more water vapor, which itself is a powerful greenhouse gas. Warmer temperatures also result in less snow and sea ice cover, allowing the

Earth to absorb more of the sun's radiation. Some simple estimates of these feedbacks increase the ECS to around 2°C (Sherwood *et al.*, 2020). Larger values of ECS are associated with positive cloud feedbacks.

Climate scientists use multiple lines of evidence to determine the Equilibrium Climate Sensitivity:

- Climate model simulations
- Historical observations
- Paleoclimatic reconstructions
- Process understanding of feedbacks

4.2 Model-based estimates of climate sensitivity

The ECS ranges given in IPCC AR4 and AR5 were obtained primarily by examining the behavior of large-scale climate models, also called General Circulation Models (GCMs). However, the IPCC changed course in its AR6 when it turned to a more direct data-driven methodology. Here we discuss some of the pitfalls of using GCMs to try to determine the Earth's climate sensitivity.

In principle, ECS is an emergent property of GCMs—that is, it is not directly parameterized or tuned, but rather emerges in the results of the simulation. Otherwise plausible GCMs and parameter selections have been discarded because of perceived conflict with an expected warming rate, or aversion to a model's climate sensitivity being outside an accepted range (Mauritsen *et al.* 2012). This practice was commonplace for the models used in AR4; modelers have moved away from this practice with time. However, even in a CMIP6 model, Mauritsen and Roeckner (2020) state the following regarding their Max Planck Institute (MPI) climate model (emphasis added):

“We have documented how we tuned the MPI-ESM1.2 global climate model to match the instrumental record of warming; an endeavor which has clearly been successful. Due to the historical order of events, the choice was to do this practically by **targeting an ECS of about 3 K using cloud feedbacks**, as opposed to tuning the aerosol forcing.”

In other words, the MPI modelers chose an ECS value of 3°C and then tuned the cloud parameterizations to match their intended result.

As noted, direct warming from CO₂ doubling is only about 1°C (Soden and Held 2006); further warming arises from climate feedbacks that are not explicitly resolved by the GCM but rely on parameterizations of physical processes. Higher values of ECS arise primarily from assumed positive cloud feedbacks, whereas the magnitude and even the sign of the feedbacks are very uncertain. Elements of cloud feedback include changes in the latitudinal distribution of clouds, changes in the distribution of cloud height (changes in low versus high clouds), changes to the phase of clouds (ice versus liquid), changes in cloud particle size (associated with changes in concentration and/or composition of aerosol particles), changes in the precipitation efficiency of clouds, and even changes in how clouds are distributed over the daily solar cycle (Curry and Webster, 1999). It is difficult for GCMs to simulate any of these processes correctly owing to their small scale, let alone predict how they will change in the future. Further, cloud processes modulate the magnitudes of the water vapor, lapse rate, and the surface albedo feedbacks.

ECS can be determined from climate model simulations by doubling the concentration of CO₂ and allowing several centuries for the warming to equilibrate. To avoid the need for such long simulations, “effective climate sensitivity” is commonly evaluated from a 150-year simulation in response to a sudden quadrupling of CO₂.

The spread of ECS values from the CMIP5 ensemble of climate models used in AR5 was 2.0–4.7°C; that range increased for the CMIP6 models used in AR6 to between 1.8 and 5.7°C (Chen *et al.*, 2021, Scaffeta 2021, see Figure 4.1). Far from resolving the model-based climate sensitivity the range appears to be growing. The main cause of the overall upward shift in ECS in CMIP6 relative to CMIP5 is a larger positive cloud feedback, driven by changes to the cloud parameterizations in many CMIP6 models (Zelinka *et al.*, 2020)

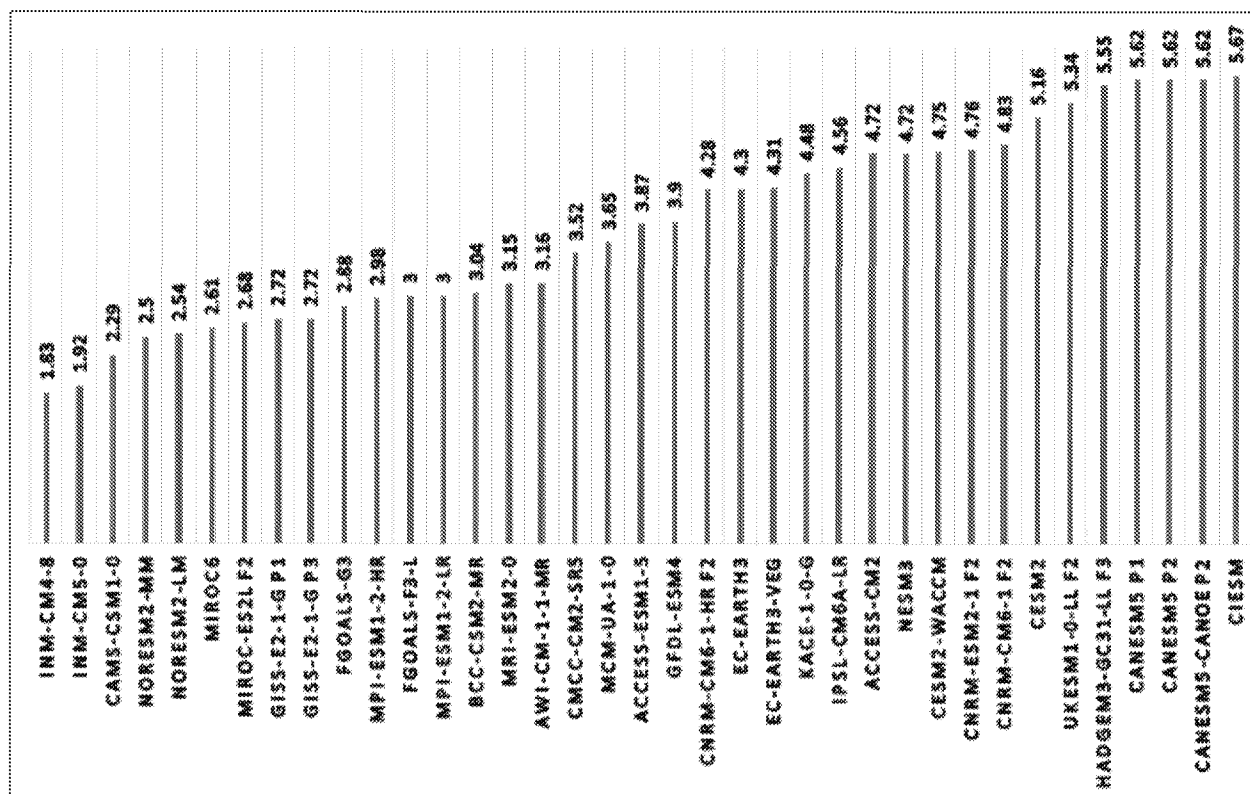


Figure 4.1 Equilibrium Climate Sensitivities in °C of 37 climate models from the CMIP6 ensemble. Identifiers for the various models appear along the horizontal axis. From (Scaffeta, 2021)

Because of concerns about model tuning and the high sensitivity to cloud parameterizations, the IPCC AR6(2021) did not rely on climate model simulations in their assessment of climate sensitivity, relying instead on data-driven methods.

4.3 Data-driven estimates of climate sensitivity

Climate sensitivity can also be estimated from instrumental records of surface temperatures and ocean heat content, combined with estimates of how climate forcings (e.g., greenhouse gases, solar, volcanoes, aerosols) have changed in the past (Otto *et al.*, 2013). Using this information, a simple empirical model called an Energy Balance Model can be employed. It requires estimating a feedback parameter whose uncertainties are highly amplified in the resulting ECS (Roe and Baker, 2007).

The accuracy of the data-driven methods depends on the quality of the input data. Assumptions are needed about ocean heat storage, and good data is only available for recent decades. The greatest source of uncertainty is the amount and composition of aerosol particles and their interactions with cloud radiative properties (the so-called aerosol indirect effect; see Figures 3.1.1, 3.1.2). Climate models exhibit warming in response to GHGs but cooling in response to aerosols (Schwartz *et al.*, 2007). Observed 20th century warming can be shown to be consistent either with low ECS and low aerosol cooling, or high ECS and high aerosol cooling. Since fossil fuel use adds both GHGs and aerosols to the atmosphere, both effects need to be estimated to isolate the warming effect of CO₂.

Paleoclimate proxies are also used to evaluate the sensitivity of past climates by comparing paleoclimate changes in the Earth's temperatures to estimates of changes in forcings. The two most informative periods are the last glacial maximum (around 20,000 years ago), which was about 3–7°C colder than today, and a mid-Pliocene period (roughly three million years ago), which was 1–3°C warmer than today. The limits on cooling during the last glacial maximum give the best single evidence that high values of climate sensitivity are unlikely. However, paleoclimate estimates are associated with very large uncertainties in the estimated temperatures and forcings. Further, estimates of climate sensitivity based on past climate states might not be applicable to the current state of the climate system.

A recurring theme in the climate literature is that ECS estimates based on historical data are smaller than ECS estimates inferred from climate models (Sherwood and Forest 2024). About 15 estimates based on historical data appeared in the peer-reviewed literature between 2012 and 2024 yielding ECS best estimates between 1.0°C and 2.5°C, although critics have questioned some of the methods and the data quality. For AR6, the IPCC placed primary weight on the results of Sherwood *et al.* (2020) that combined historical data and paleoclimate proxies with the process-based approach and yielded a best estimate of 3.1°C with a *likely* range of 2.6–3.9°C. Lewis (2022) raised a number of concerns about this result, including methodological errors, outdated input values, and use of subjective Bayesian priors in the analysis. Lewis' analysis found that climate sensitivity is estimated to be much lower and better constrained than in the Sherwood *et al.* analysis – median 2.2°C (1.8–2.7°C in the 17–83 percent *likely* range, and 1.6–3.2°C in the 5–95 percent *very likely* range). The IPCC AR6 estimated only a 5 percent probability that ECS was below 2.3°C, whereas Lewis estimated it to be over 50 percent. The most recent publications on the debate between Sherwood *et al.* and Lewis further defend their respective positions: Sherwood and Foster (2024) and Lewis (2025).

An argument emphasized in AR6 is that data-driven ECS estimates might understate the future warming response to GHGs because of a so-called “pattern effect” (Forster *et al.*, 2021). The tropical Pacific is believed to strongly influence the overall efficiency with which the Earth radiates heat to space, but some regions remove heat more efficiently than others. If the west-to-east temperature gradient in the tropical Pacific is weakened in a warming climate, warming would concentrate where heat is less efficiently removed, raising ECS.

Most climate models simulate that rising GHGs will weaken the west-east temperature gradient, which led the IPCC in AR6 to conclude that data-driven ECS estimates understated the *future* ECS value. However, Seager *et al.* (2019) pointed out that, contrary to models, the west-east temperature gradient has been strengthening over time. They further argued that the mechanism predicting otherwise in climate models was based on a faulty characterization of oceanic dynamics and there is no reason to expect the gradient to weaken. A similar argument was recently made by Lee *et al.* (2024), who concluded that “the trajectory of the observed trend reflects the response to increasing GHG loading in the atmosphere”; in other words, GHG warming should lead to a future strengthening rather than a weakening of the temperature gradient. Increased efficiency of atmospheric cooling implies, if anything, that the future ECS in a warming climate might be *lower* than current estimates.

4.4 Transient Climate Response

The Transient Climate Response (TCR) provides a more useful observational constraint on climate sensitivity. TCR is the global temperature increase that results when CO₂ is increased at an annual rate of 1 percent over a period of 70 years (*i.e.*, doubled gradually). Relative to the ECS, observationally determined values of TCR avoid the problems of uncertainties in ocean heat uptake and the fuzzy boundary in defining equilibrium arising from a range of timescales for the longer-term feedback processes (*e.g.*, ice sheets). TCR is more generally related to peak warming and better constrained by historical warming, than ECS. IPCC AR6 judged the *very likely* range of TCR to be 1.2–2.4°C. In contrast to ECS, the upper bound of TCR is more tightly constrained. For comparison, the TCR values determined by Lewis (2023) are 1.25 to 2.0°C, showing much better agreement with AR6 values than was seen in a comparison of the ECS values.

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5 MODELS VERSUS OBSERVATIONS IN THE RECENT PAST

Chapter summary

Climate models show warming biases in many aspects of their reproduction of the past several decades. In response to estimated changes in forcing they produce too much warming at the surface (except in the models with lowest ECS), too much warming in the lower-and mid-troposphere and too much amplification of warming aloft.

Climate models also produce too much recent stratospheric cooling, invalid hemispheric albedos, too much snow loss, and too much warming in the Corn Belt. The IPCC has acknowledged some of these issues but not all.

5.1 Introduction

Climate models are the primary tool used to project future climate changes in response to increasing atmospheric levels of anthropogenic greenhouse gases. To assess the fitness of climate models for this purpose, it is reasonable to ask how well they reproduce the current climate and its variations over the past century. The box “Climate modeling” gives some detail on how climate models work.

Of great concern is the fact that, after several decades of the climate modeling enterprise involving approximately three dozen models operated by research centers around the world, the range of future warming they produce in response to a hypothetical doubling of atmospheric CO₂ extends over a factor of three, as we discussed in the previous chapter. This range of disagreement among models has not decreased for decades.

Problems with climate models are not just in their disagreement over the future, but also in their ability to replicate the recent past. Here we review some of the most important metrics of climate model accuracy: ability to reproduce historical surface, tropospheric and stratospheric temperature trends; ability to reproduce the vertical warming profile; and ability to reproduce other climate features such as snowfall. In all cases a persistent finding is that models on average err on the side of too much warming in response to estimated historical forcings.

BOX: Climate modeling

Most climate models, except for the simplest ones, represent Earth's land surface using a grid of squares on the order of 100 km wide. To simulate the atmosphere, typically 30 or more grid boxes are stacked above these squares. The ocean is modeled using a similar but finer grid, resulting in tens of millions of grid boxes for the atmosphere and oceans.

The computer models, based on physical laws, calculate how air, water, and energy move between grid boxes over time. The time step can be as small as 10 minutes, and repeating this process millions of times allows the simulation of climate over centuries. Running these models, even on the most powerful supercomputers, can take months. Comparing simulation results with historical climate data helps assess the accuracy of a model, while projections into the future estimate climate changes under assumed human and natural influences.

Despite sounding straightforward, climate modeling is highly complex. Many critical processes occur at scales smaller than the grid size. For instance, the flows of sunlight and heat in the atmosphere depend strongly on cloud cover. Since tracking individual clouds is impractical, researchers must make “subgrid” assumptions about the distribution of clouds in each grid box. Snow and ice cover, which affects how much sunlight is reflected or absorbed by the surface, is another subgrid factor.

Each subgrid assumption requires numerical parameters, which must be carefully set. Modelers initially estimate these parameters based on physics and observed climate patterns, then run the model. Because early results often diverge significantly from real-world observations, they “tune” these parameters to better match observed climate features. Different modeling teams use distinct assumptions and tuning strategies, leading to varied outcomes. Tuning is a necessary but delicate aspect of climate modeling, as it is for any complex system. Poor tuning can result in inaccurate simulations, while excessive tuning risks artificially steering results toward predetermined conclusions.

The spread of model representations of the current climate is very wide. One of the most basic indicators—Earth's average surface temperature—varies by about 3°C across CMIP6 models prior to 1880 (Figure 5.1), narrows slightly until 2040 then diverges to over 4 °C. For comparison 20th century warming was only about 1.0°C. This variation suggests substantial differences among models' physical processes.

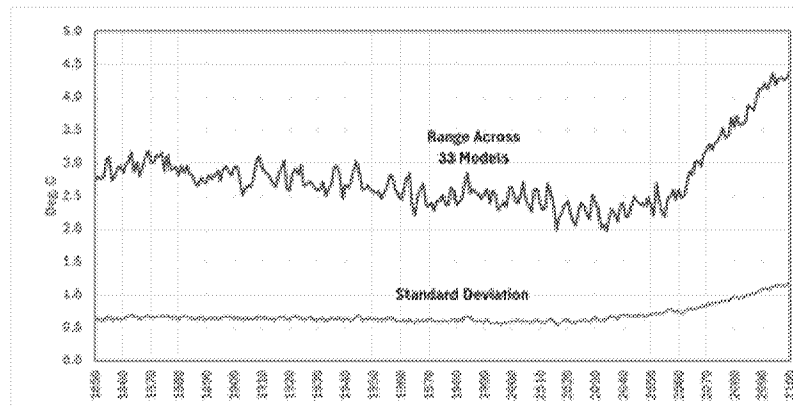


Figure 5.1 CMIP6 Average surface temperature range across 33 models and standard deviation using SSP5-85 scenario. Data from KNMI Climate Explorer website at [[HYPERLINK "https://climexp.knmi.nl/start.cgi" \t "_blank"](https://climexp.knmi.nl/start.cgi?_blank)].

Beyond the models' ability to reproduce features of today's climate, the critical issue for society is how well they predict responses to subtle human influences, such as greenhouse gas emissions, aerosol cooling, and land-use changes. The most crucial aspect that models must capture correctly is “feedbacks.” These occur when climate changes either amplify or suppress further warming. In general, the modeled net effect of all feedbacks doubles or triples the direct warming impact of CO₂.

5.2 Surface warming

A straightforward test of a climate model's validity is its ability to reproduce historical warming in response to known past changes in climate drivers such as greenhouse gases. Figure 5.2 is reproduced from Scafetta (2023), that groups the latest-generation (CMIP6) climate models into low ECS (1.5 to 3.0 degrees C), medium ECS (3.0 to 4.5 degrees C) and high ECS (4.5 to 6.0 degrees C), and compares their post-1980 global average temperature simulation ranges to those of three surface temperature records and one satellite-based lower troposphere temperature data product.

The leftmost column shows that the low-ECS models track the post-1980 historical warming record reasonably well, but the middle and right columns show that the medium- and high-ECS models conspicuously over-predict the warming.

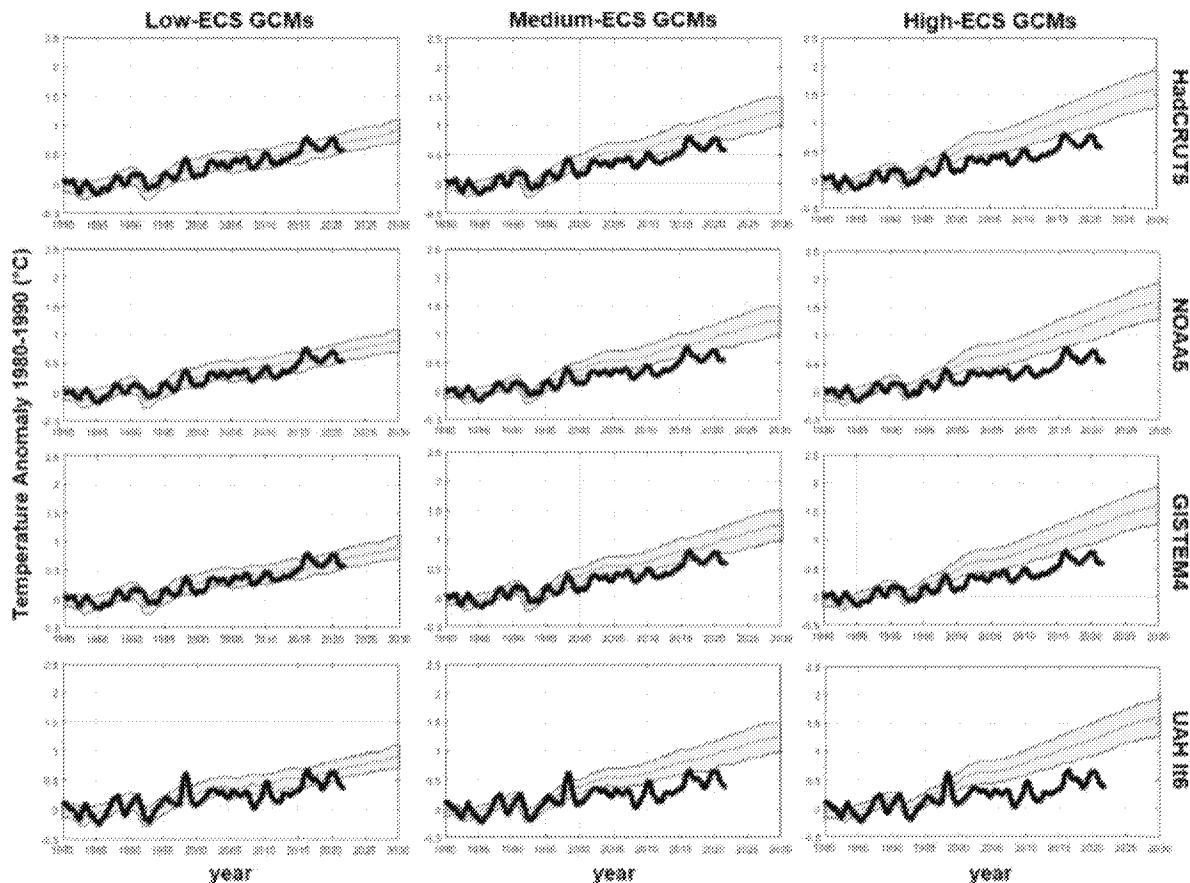


Figure 5.2: Model-Observational comparisons for Earth's surface warming. The columns correspond to model groups showing low-ECS to high-ECS, while the rows correspond to widely-used observed temperature records, the first three showing surface averages and the fourth showing the lower troposphere average. In each panel the yellow area denotes the mean and range ($\pm$ one standard deviation) of climate model simulations for that group. The thick black line shows the observed annual average temperature in the indicated source. Source: Scafetta (2023) Fig.2.

Spencer (2024) has also provided a useful summary of the Model-Observation mismatch by comparing trends in surface temperature data products with those in individual climate models, as summarized in Figure 5.3; most climate models show substantially more warming than the observations since 1979.

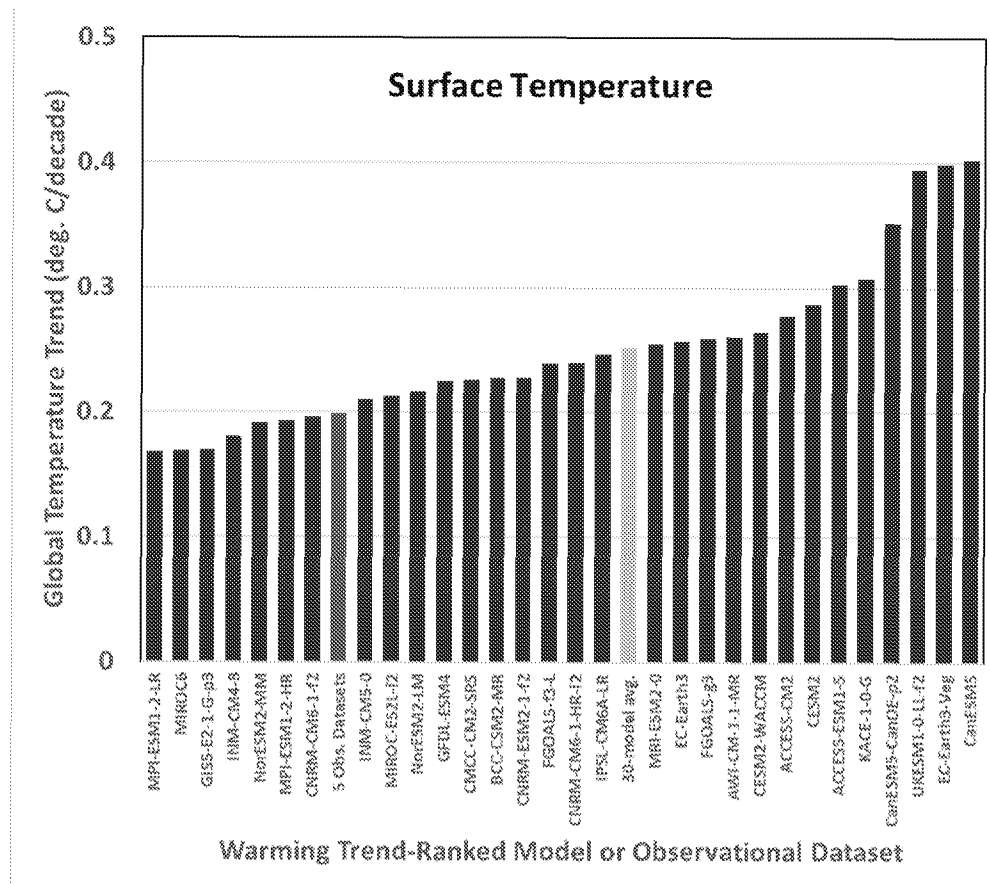


Figure 5.3 Global surface air temperature trends ($^{\circ}\text{C}/\text{decade}$), 1979-2024, from various CMIP6 climate models (red, 30-model average in orange); and the average of three thermometer datasets (HadCRUT5, NOAA Global Temp, and Berkeley 1 deg.) and two reanalysis datasets (ERA5 and NCEP/NCAR R1) in blue. Data source: <https://climexp.knmi.nl/start.cgi>.

5.3 Tropospheric warming

It has long been known that climate models on average overstate warming in the tropical troposphere. This region is an important test of climate models since this is where the signal of anthropogenic greenhouse warming emerges first and most strongly. Biases in tropospheric trends indicate model flaws in heat transfer processes that carry over to surface warming biases.

The discrepancy was flagged as a serious inconsistency in the first U.S. Climate Change Science Program report (Karl *et al.* 2006) and has been mentioned in every IPCC report since, but the discrepancy has gotten worse over time, and the bias is now global. McKittrick and Christy (2020) compared tropospheric warming trends in CMIP6 climate models to observed trends from satellites, weather balloons and reanalysis systems. Every model overpredicted the average observed warming trend over 1979-2014 in both Lower- and Mid-Troposphere layers, both globally and in the tropics. In most individual models the bias was statistically significant and on average across models it was highly significant.

Figure 5.4 presents the comparisons with data updated to 2024 (McKittrick and Christy 2025). The recent warm years moved the observed trend up slightly and widened the trend confidence intervals but the overall pattern remains the same: model bias is towards too much warming, in most cases the difference is statistically significant and on average the bias is statistically highly significant. McKittrick and Christy (2020) also showed that the bias is larger in high-ECS models, but even the models with lower average ECS predict too much warming. If future climate models were to realistically represent global tropospheric warming, they would likely be less sensitive than even the low-ECS members of the CMIP6 ensemble.

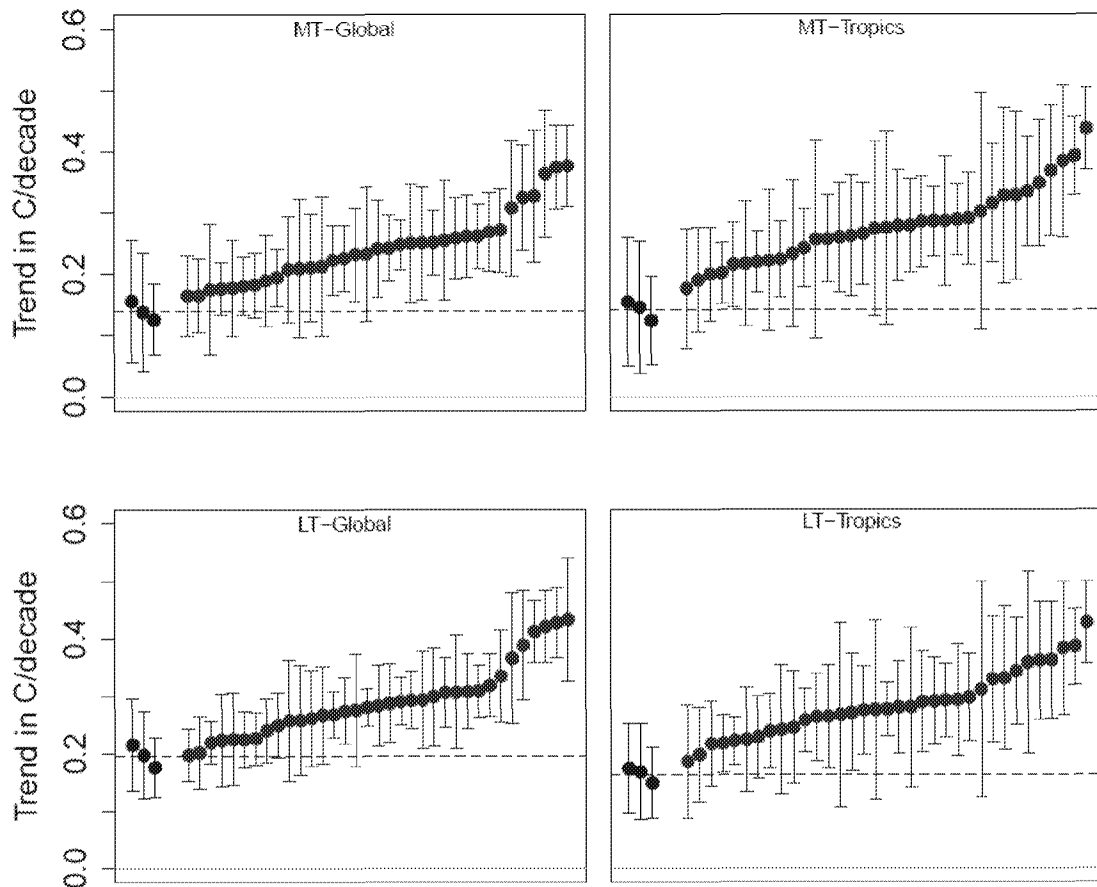


Figure 5.4: Observed versus CMIP6 modeled warming trends (°C/decade 1979-2024) in the global and tropical lower (LT) and mid-troposphere (MT). Uses methodology of McKittrick and Christy (2020) on data updated from 2014 to 2024. Blue dots: warming trends with 95 percent confidence intervals for 3 data products (radiosondes, reanalysis, and satellites). Blue dashed line: warming trend average for 3 observed series. Red dots: modeled warming trends with 95 percent confidence intervals in 35 models arranged lowest to highest.

As mentioned previously, the IPCC has long acknowledged the model-observation mismatch. For example, AR6 pp. 443-444 offers this on the tropical troposphere (it does not address the global comparison):

Several studies since AR5 have continued to demonstrate an inconsistency between simulated and observed temperature trends in the tropical troposphere, with models simulating more warming than observations (Mitchell *et al.*, 2013, 2020; Santer *et al.*, 2017a, b; McKittrick and Christy, 2018; Po-Chedley *et al.*, 2021)... Over the 1979–2014 period, models are more consistent with observations in the lower troposphere, and least consistent in the upper troposphere around 200 hPa, where biases exceed 0.1°C per decade. Several studies using CMIP6 models suggest that differences in climate sensitivity may be an important factor contributing to the discrepancy between the simulated and observed tropospheric temperature trends (McKittrick and Christy, 2020; Po-Chedley *et al.*, 2021), though it is difficult to deconvolve the influence of climate sensitivity, changes in aerosol forcing and internal variability in contributing to tropospheric warming biases (Po-Chedley *et al.*, 2021). Another study found that the absence of a hypothesized negative tropical cloud feedback could explain half of the upper troposphere warming bias in one model (Mauritsen and Stevens, 2015).

... In summary, studies continue to find that CMIP5 and CMIP6 model simulations warm more than observations in the tropical mid- and upper-troposphere over the 1979–2014 period (Mitchell *et al.*, 2013, 2020; Santer *et al.*, 2017a, b; Suárez-Gutiérrez *et al.*, 2017; McKittrick and Christy, 2018), and that overestimated surface warming is partially responsible (Mitchell *et al.*, 2013; Po-Chedley *et al.*, 2021). Hence, we assess with *medium confidence* that CMIP5 and CMIP6 models continue to overestimate observed warming in the upper tropical troposphere over the 1979–2014 period by at least 0.1°C per decade,

Notably, despite the accumulation of evidence of excess model warming the IPCC assigns only *medium confidence* to the existence of a warming bias.

5.4 Vertical temperature profile mismatch

Another important model-observational discrepancy is the excess amplification with altitude found in climate models. The comparison was in the IPCC AR5 Chapter 10, although only in the online supplement (Figure 10.SM.1) and only in a figure whose formatting obscured the point. Figure 10.SM.1 is not referenced in the main IPCC report nor in any summary so readers would not have been aware of it. Although not apparent at first glance, it shows that the 1979-2010 warming in the lower troposphere is so small as to be consistent with no GHG forcing at all and is inconsistent with the model runs that do have GHG forcing. In Figure 5.6 we adapt IPCC AR5 Figure 10.SM.1 to draw out this critical point.

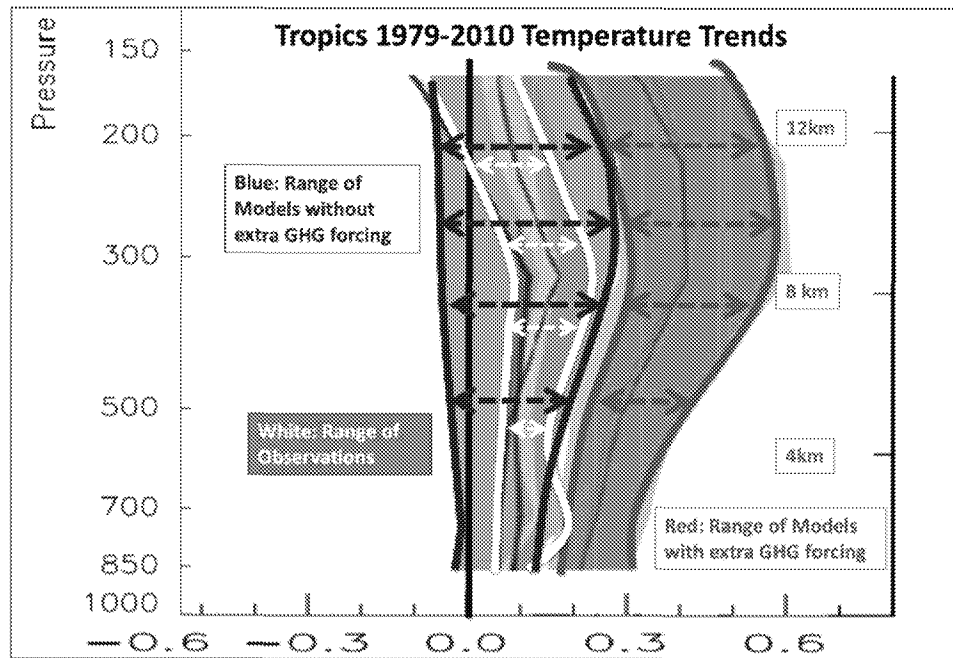


Figure 5.6: Vertical warming pattern for tropics (20S to 20N). Source: Annotated version of IPCC AR5 Figure 10.SM.1

Figure 5.6 compares model and observational temperature trends by altitude between 20S and 20N (the tropics). In this region where the models say the warming should be strongest, the observations (shown here in white) lie within the blue “No CO₂” band and entirely outside the “with CO₂” red envelope. This means that in the entire tropical atmospheric column from the surface to the base of the stratosphere, observed warming trends are so small as to be consistent with the output of models that have no anthropogenic CO₂, and inconsistent with the entire envelope of warming trends generated by models forced with increased CO₂.

A similar comparison is shown in Christy and McNider (2017), an updated version of which (covering 1979-2024) is reproduced as Figure 5.7. Modeled temperature trends exceed observations from the surface through the top of the troposphere, with observed trends below the entire model range at most pressure levels. Also shown in Figure 5.7 is the tropical tropospheric temperature (TTT) average from three satellite data products (NOAA, UAH and RSS) compared to the same layer average from climate models for 1979-2024. Again, the observed trends lie below the entire model range.

The wide range of choices made by modelers to characterize the physical processes in the models (see Box: Climate Modeling in Section 5.1 above) is seen by the large spread of trends in the middle troposphere, ± 40 percent about the median (Fig. 5.7). This vividly illustrates the uncertainties in attempts to model (parameterize) a complex system involving turbulence, moist thermodynamics, and energy fluxes over the full range of the tropical atmosphere’s time and space scales.

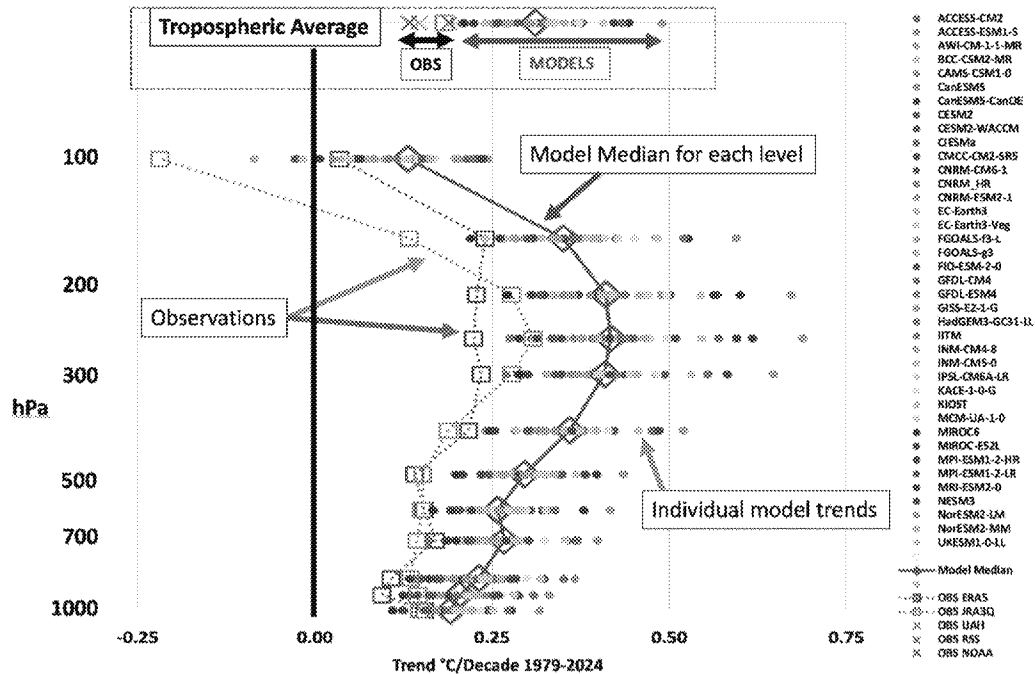


Figure 5.7: Modeled versus observed warming, tropical troposphere. Source: updated from Christy and McNider (2017) including data through 2024. Red line: model average. Green and blue lines: observational series (reanalysis).

This discrepancy has been the source of much controversy, with some arguing that even if there is very little observed warming aloft in the tropics, a “hotspot” still exists in the sense that the warming aloft is greater than at the surface (Santer et al. 2008). But there is good evidence that models also exaggerate the amplification rate. Klotzbach *et al.* (2009), showed that models project greater amplification with altitude than is observed. This result was subsequently confirmed by detailed time series analysis (Vogelsang and Nawaz 2016) which found that the model-observational difference is statistically significant.

The atmosphere’s temperature profile is a case where models are not merely uncertain but also show a common warming bias relative to observations. This suggests that they misrepresent certain fundamental feedback processes.

The IPCC AR6 did not assess this issue.

5.5 Stratospheric cooling

An important element of the expected general “fingerprint” of anthropogenic climate change is simultaneous warming of the troposphere and cooling of the stratosphere. AR6 acknowledged that cooling had been observed but only until the year 2000. The stratosphere has shown some warming since, contrary to model projections.

AR6 WG1 Ch 2 pp. 327-9 states:

Temperatures averaged through the full lower stratosphere (roughly 10–25 km) have decreased over 1980–2019 in all data products, with the bulk of the decrease prior to 2000. The decrease holds even if the influence of the El Chichon (1982) and Pinatubo (1991) volcanic eruptions on the trend,

found by Steiner *et al.* (2020a) to have increased the 1979–2018 cooling trend by 0.06°C per decade, is removed. Most datasets show no significant or only marginally significant trends over 2000–2019, and the results of Philipona *et al.* (2018) show weak increases over 2000–2015 in the very lowermost stratosphere sampled by radiosondes....

It is virtually certain that the lower stratosphere has cooled since the mid-20th century. However, most datasets show that lower stratospheric temperatures have stabilized since the mid-1990s with no significant change over the last 20 years. It is likely that middle and upper stratospheric temperatures have decreased since 1980, but there is low confidence in the magnitude.

The cited source, Philipona *et al.* (2018), in an article entitled “Radiosondes Show That After Decades of Cooling, the Lower Stratosphere Is Now Warming”, stated:

In response to continued greenhouse gas increases and stratospheric ozone depletion, climate models project continued tropospheric warming and stratospheric cooling over the coming decades. Global average satellite observations of lower stratospheric temperatures exhibit no significant trends since the turn of the century. In contrast, an analysis of vertically resolved radiosonde measurements from 60 stations shows an increase of lower stratospheric temperature since the turn of the century at altitudes between 15 and 30 km and over most continents. Trend estimates are somewhat sensitive to homogeneity assessment choices, but all investigated radiosonde data sets suggest a change from late twentieth century cooling to early 21st century warming in the lower stratosphere.

A combination of tropospheric warming and stratospheric cooling is a commonly cited “fingerprint” of anthropogenic warming. Stratospheric warming since 2000 coincides with continued surface and tropospheric warming, a pattern that is not found in climate model simulations and is not apparently consistent with the anthropogenic fingerprint.

5.6 Snow cover mismatch

Data compiled by the Rutgers University Snow Lab show that Northern Hemisphere winter snow cover is not decreasing (Figure 5.8); if anything, it shows an increasing trend.

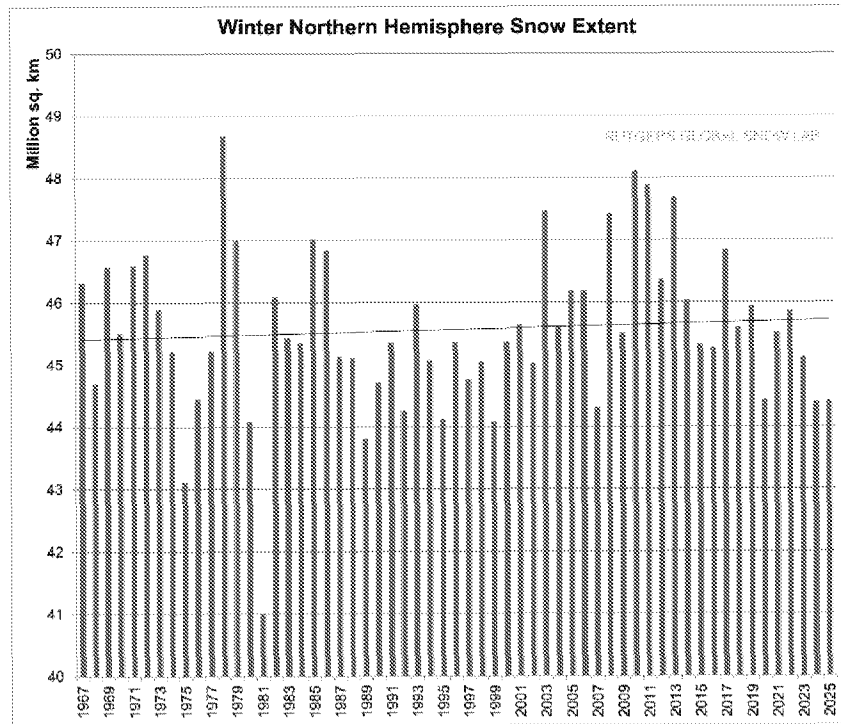


Figure 5.8: Northern Hemisphere Winter Snow extent.

Source: [[HYPERLINK](https://climate.rutgers.edu/snowcover/chart_seasonal.php?ui_set=nhland&ui_season=1)]

"https://climate.rutgers.edu/snowcover/chart_seasonal.php?ui_set=nhland&ui_season=1"]
(accessed May 27, 2025)

Yet models project declining Northern Hemisphere snow cover in a warming climate, as described by Connolly *et al.* (2019).

The climate models were found to poorly explain the observed trends [in Northern Hemisphere snow cover]. While the models suggest snow cover should have steadily decreased for all four seasons, only spring and summer exhibited a long-term decrease, and the pattern of the observed decreases for these seasons was quite different from the modelled predictions. Moreover, the observed trends for autumn and winter suggest a long-term increase, although these trends were not statistically significant.

AR6 largely confines its discussion of changing Northern Hemisphere snow cover extent (SCE) to the Spring season, for which models and observations agree on a downward trend. Regarding Winter changes it remarks as follows (AR6 WGI Ch. 2 p. 344):

Assessment of SCE trends in the NH since 1978 indicates that for the October to February period there is substantial uncertainty in trends with the sign dependent on the observational product. Analysis using the NOAA Climate Data Record shows an increase in October to February SCE (Hernández-Henríquez *et al.*, 2015; Kunkel *et al.*, 2016) while analyses based on satellite borne optical sensors (Hori *et al.*, 2017) or multi-observation products (Mudryk *et al.*, 2020) show a negative trend for all seasons.

AR6 WGI Chapter 9 (p. 1284) points out that the NOAA Climate Data Record showing increased fall and winter SCE is inconsistent with land-based observations and model-based data sets. It notes that using optical satellite imagery to infer SCE in winter is challenging due to cloud cover and decreased solar illumination in winter months. Focusing on the Pacific Coast states (CA, OR and WA) cold season mountain snowfall that melts in spring and summer provides a substantial portion of warm season water resources. A comprehensive reconstruction of snowfall for the main source regions (Cascades and Sierra Nevada Mountains) indicates no significant trends in annual totals since the late 19th century (Christy 2022).

In summary, the up-to-date Rutgers SCE database indicates a mismatch between models and observations. Further work to reconcile conflicting trends in observational data sets is required.

5.7 Hemispheric symmetry of the planetary albedo

The planetary albedo is the fraction of incoming solar radiation reflected by the Earth back to space. It is an important element of the radiative energy balance and influences whether the planet will warm or cool over time. The planetary albedo is typically estimated at around 0.30; variations on the scale of 0.01 correspond to changes in solar forcing of about 3 W/m^2 , an amount larger than current anthropogenic forcing. It has long been noted that models disagree with each other and with observations on the value of the global planetary albedo (Stephens *et al.* 2015).

An intriguing property of the Earth's albedo is that, on average, the Northern Hemisphere (NH) and Southern Hemisphere (SH) have had nearly the same albedo, at least throughout the fifty-year satellite record (Stephens *et al.*, 2015). This symmetry is surprising, because the SH has much more ocean than land. Since ocean is less reflective than land, the NH should have higher albedo. Clouds (which are highly reflective) are more common in the NH and so compensate the surface albedo imbalances of the two hemispheres. Datseris and Stephens (2021) show that this cloud compensation comes from the extra-tropical storm tracks of the SH, which are cloudier than those of the NH. While the mechanism for this hemispheric symmetry is unclear, it likely operates on large temporal and spatial scales.

The hemispheric symmetry of the albedo is a simple gross metric for climate models. Rugenstein and Hakuba (2023) defined that metric as the difference between the NH and SH annual mean albedos, expressed as Wm^{-2} of the reflected sunlight, and compiled it for the CMIP6 climate models, as shown in Figure 5.9. Most of the CMIP6 models do not reproduce the small observed asymmetry (about 0.1 Wm^{-2}) and even disagree as to which hemisphere is more reflective. Moreover, the magnitude of the asymmetry ranges up to 5 Wm^{-2} in some models, twice as large as the current anthropogenic forcing (about 2.5 Wm^{-2}).

The significance of unphysical albedo asymmetries in the climate models is not yet fully known. However, other model studies suggest that interhemispheric changes in albedo can alter poleward heat fluxes, meridional temperature gradients, storminess, and differences in hemispheric ocean heat storage. The discrepancy between models and observations further raises issues regarding cloud feedback processes and so more generally diminishes confidence in model projections of the future climate.

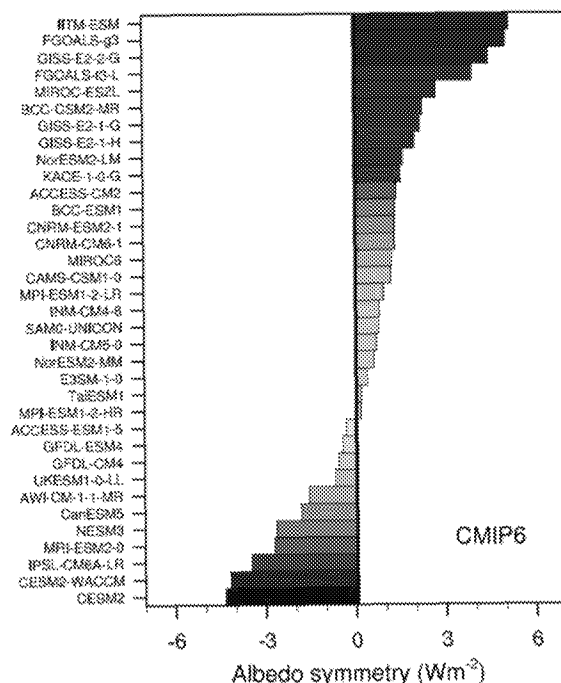


Figure 5.9. Differences in 20-year average reflectivity (albedo) between the Northern and Southern Hemispheres for CMIP6 models used in the most recent IPCC assessment (colored bars). The very small observed difference is indicated by the vertical black line. From Rugenstein and Hakuba (2023).

5.8 U.S. Corn Belt

One of the largest discrepancies between models and observations is in the U.S. Corn Belt, a region of particular importance to global food production. Figure 5.10 shows the warming trends for summertime (June, July, August) for the 12-state Corn Belt (IN, IA, IL, ND, SD, MO, MN, WI, MI, OH, KS, NE) during 1973-2022. All 36 climate models (red) warm far too rapidly compared to observations (blue).

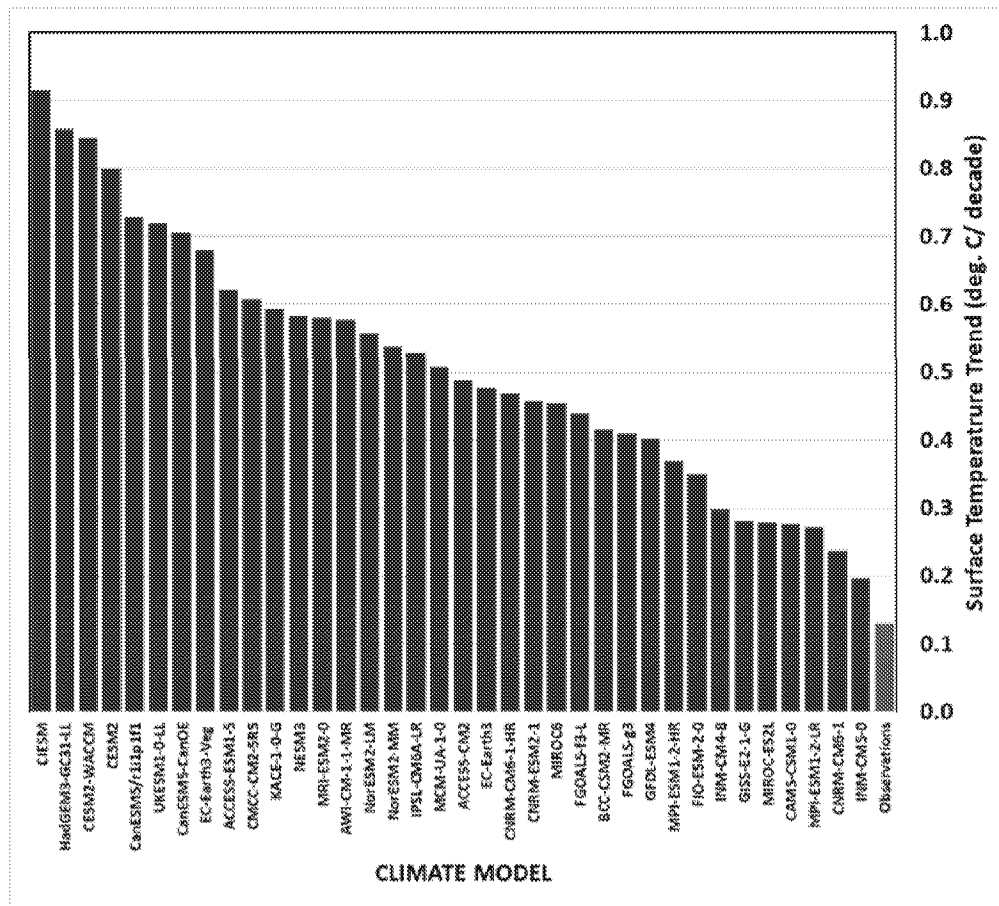


Figure 5.10: Modeled versus observed warming trends in the U.S. Corn Belt, 1973-2022.

As discussed in Chapter 9, the anticipated negative effects of increasing temperatures on U.S. corn yields have not materialized, in contrast to widely publicized studies proclaiming that theoretical future impacts are already being experienced (*e.g.*, Seager *et al.*, 2018).

The IPCC acknowledges limitations in the accuracy of regional climate model outputs. This example shows that users need to assess model projections carefully on a case-by-case basis since local biases might be sufficiently large that the models are simply not fit for purpose. As has recently been noted by two leaders of the modeling community (emphasis added)

... for many key applications that require regional climate model output or for assessing large-scale changes from small scale processes, we believe that **the current generation of models is not fit for purpose**. (Palmer and Stevens 2019)

To summarize:

- Climate models show warming biases in many aspects of their reproduction of the past few decades.
- They produce too much warming at the surface (except in the models with lowest ECS), too much warming in the lower-and mid-troposphere and too much amplification of warming aloft
- They also produce too much stratospheric cooling, too much snow loss, and too much warming in the U.S. Corn Belt.

- The hemispheric albedo difference in individual climate models ranges widely in sign and magnitude compared to observations. The range in W/m^2 is three times larger than the direct anthropogenic forcing of CO_2 .
- The IPCC has acknowledged some of these issues but not others.

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6 EXTREME WEATHER

Chapter Summary

Most types of extreme weather exhibit no statistically significant long-term trends over the available historical record. While there has been an increase in hot days in the U.S. since the 1950s, a point emphasized by AR6, numbers are still low relative to the 1920s and 1930s. Extreme convective storms, hurricanes, tornadoes, floods and droughts exhibit considerable natural variability, but long-term increases are not detected. Some increases in extreme precipitation events can be detected in some regions over short intervals but the trends do not persist over long periods and at the regional scale. Wildfires are not more common in the U.S. than they were in the 1980s. Burned area increased from the 1960s to the early 2000's, however it is low compared to the estimated natural baseline level. U.S. wildfire activity is strongly affected by forest management practices.

6.1 Introduction

High impact weather extremes, usually related to temperature, precipitation and/or high wind, can disrupt infrastructure and therefore endanger human health and wellbeing. The issue is not whether extremes occur. Rather, it is whether there are long-term (decadal scale) changes in the frequency or character of extremes (“detection”), as well as the extent to which such changes and the attendant changes in hazards are caused by anthropogenic emissions of greenhouse gases (“attribution”; e.g., AR6 Seneviratne *et al.* 2021).

It is naïve to assume that any recent extreme event is caused by human influences on the climate. Climate is about the statistical properties of weather over decades, not single events. Further, there are only about 130 years of reliable observational records that can be analyzed statistically. That brief interval does not begin to contain all the extreme events that the climate system can create on its own. Over geologic time the climate system has generated an (essentially) infinite variety of weather patterns and extremes that humans have never observed and thus are absent from the databases used to determine extreme statistics [see *Perils of short data records* below]. For that reason, attributing an extreme event unprecedented in the record requires assumptions about the magnitude of natural variations.

This chapter is concerned with detection of trends in extreme weather, while Chapter 8 considers causal attribution, with Section 8.4 specifically addressing extreme weather. If no trend is detected, then clearly there is no basis for attribution. But even where a trend is observed, attribution to human-caused warming does not necessarily follow.

This is especially true for precipitation events. The hydrological literature has long noted the phenomenon of Long-Term Persistence (LTP) in rainfall data (Hurst 1951, Cohn and Lins 2005, Markonis and Koutsoyiannis 2016). LTP has a formal mathematical definition, but it can be thought of as the presence of natural variations with strong autocorrelations that result in cycle-like behavior on long time scales. The presence of LTP requires long records to accurately estimate variability. Analysis of records that are short relative to the LTP scale will tend to underestimate variability, therefore overstating the significance of apparent trends and underestimating the likelihood of extreme events (Cohn and Lins 2005).

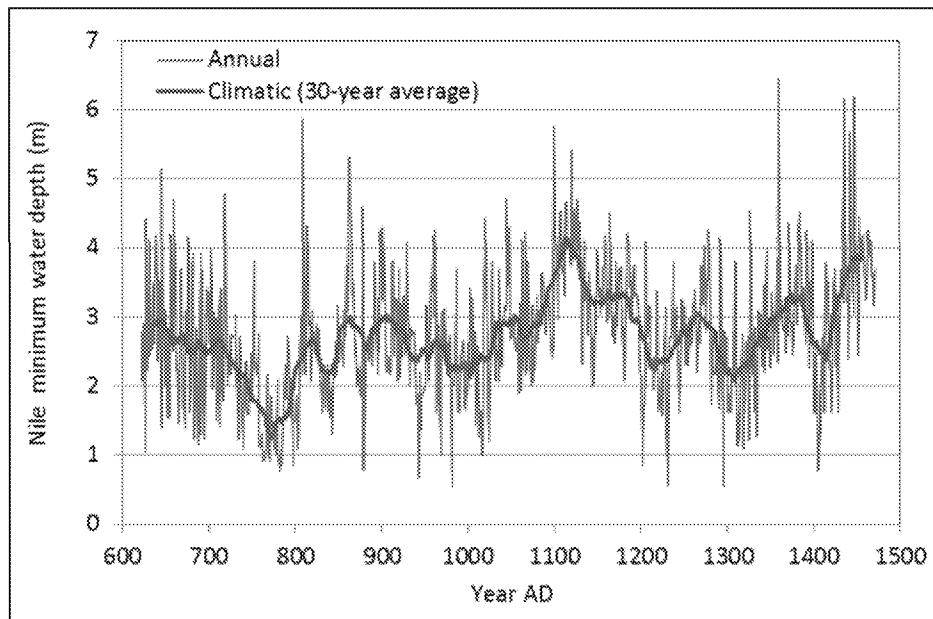


Figure 6.1.1: The annual minimum depth of the Nile River near Cairo over more than 650 years from 622 to 1284 A.D. The data, measured in meters, shows a characteristic pattern of year-to-year fluctuations around longer-term trends. Data from Koutsoyiannis (2013)

A good example of LTP is the eight-century long record of the annual minimum height of the Nile River observed at Roda Island in Cairo shown in Figure 6.1.1. The Nile River is fed by precipitation over a 4 million square mile drainage basin, an area equal to about one third of CONUS. Since human influences on the global climate were negligible before the 20th century, the century-scale variability of the thirty-year average is entirely natural; Egyptians of the seventh and eighth centuries would have been incorrect to assume that the worsening drought during that time was the “new normal.”

With these caveats in mind, we examine the evidence for changes in selected weather and climate extremes. A recurring theme is the wide gap between public perceptions and scientific evidence. It has become routine in media coverage, government and private sector discussions, and even in some academic literature to make generalized assertions that extreme weather of all types is getting worse due to GHGs and “climate change.” Yet expert assessments typically have not drawn such sweeping conclusions and instead have emphasized the difficulty both of identifying specific trends and establishing a causal connection with anthropogenic forcing.

In the sections to follow we provide excerpts from various IPCC and NCA assessment reports denoting the sources as follows:

SREX: The IPCC Special Report on Managing the Risks of Extreme Events and Disasters to Advance Climate Change Adaptation (2012)

AR6: The IPCC Sixth Assessment Report Working Group 1 (2021).

NCA4: The U.S. Climate Science Special Report of the Fourth National Climate Assessment (2017) Volume I.

NCA5: The U.S. Climate Science Special Report of the Fifth National Climate Assessment (2023) Volume I.

In the excerpts, *italics* are in the original whereas **boldface** emphasis has been added.

Additionally we present updated data series from standard government sources to provide up-to-date information through 2024 wherever possible.

6.2 Hurricanes and tropical cyclones

AR6 provides the following assessment of tropical cyclones (TCs; used here as a synonym for hurricanes):

AR6: There is *low confidence* in most reported long-term (multidecadal to centennial) trends in TC frequency or intensity-based metrics due to changes in the technology used to collect the best-track data. (IPCC, 2021 p. 1585)

AR6: It is *likely* that the global proportion of major (Category 3–5) tropical cyclone occurrence has increased over the last four decades . . . There is *low confidence* in long-term (multi-decadal to centennial) trends in the frequency of all-category tropical cyclones. (IPCC, 2023 SPM p. 9)

AR6: A subset of the best-track data corresponding to hurricanes that have directly impacted the United States since 1900 is considered to be reliable, and shows **no trend in the frequency of U.S. landfall events**. (IPCC 2021 p. 1585)

Since 1980, when satellite observations first fully covered the global oceans, we have confidence in the numbers of total global hurricanes and major hurricanes (Category 3 and higher). Figure 6.2.1 shows that on average, each year there are about 50 hurricanes, with about 25 reaching major hurricane status (Maue, 2025). There is substantial year-to-year and decadal variability, a weak decrease in the number of hurricanes, and a slight but insignificant increase in the number of major hurricanes. These two trends combine to create an overall increase in the proportion of major hurricanes.

Global hurricane statistics are dominated by the Northwest Pacific Ocean, which accounts for ~35 percent of total global hurricanes, whereas the Atlantic accounts for ~15 percent of global hurricanes (Colorado State University, 2025). Data in the Atlantic basin extends further back than in the other ocean basins and is most relevant to U.S. policy makers.

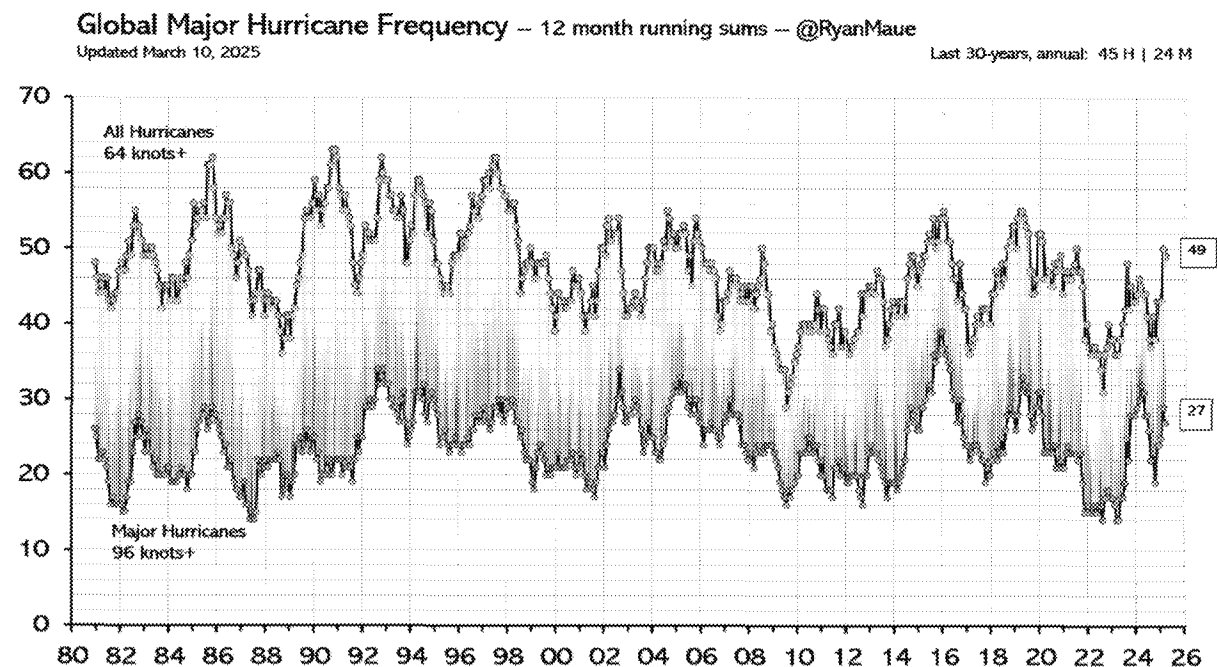


Figure 6.2.1: Global frequency of hurricanes and major hurricanes since 1980.

Figure 6.2.2 shows the frequency of Atlantic hurricanes and major hurricanes (Category 3 and higher) back to 1920. Data prior to 1965 (the onset of satellite observations in the Atlantic, shaded in blue) shows some undercounting, with data prior to 1920 showing substantial undercounting (Vecchi and Knutson, 2011). All measures of Atlantic hurricane activity show a significant increase since 1970. However, the period from 1971-1994 saw exceptionally low activity, with high activity (comparable to the past two decades, even with undercounting) also observed during the 1950's and 1960's, and even in the 1930's.

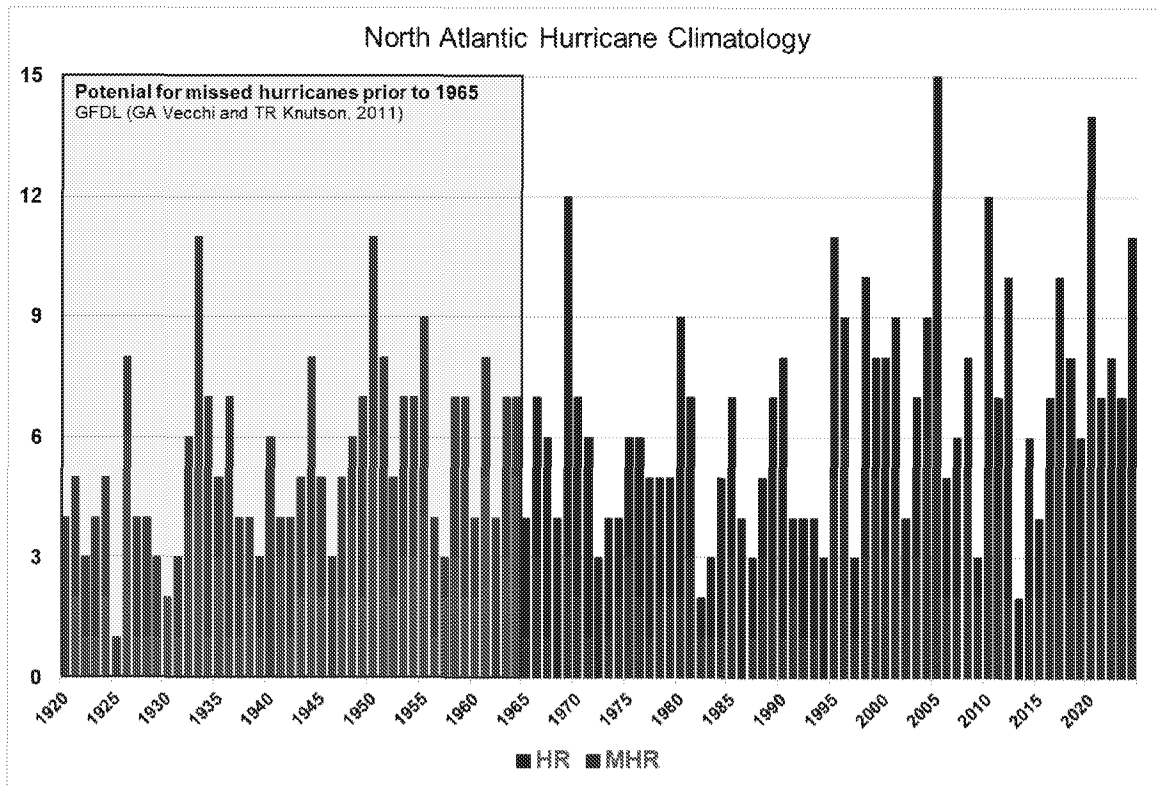


Figure 6.2.2: Atlantic frequency of hurricanes (HR) and major hurricanes (MHR) since 1920. Source National Hurricane Center (2024)

Figure 6.2.2 shows that Atlantic hurricanes vary strongly on decadal and multidecadal time scales. These variations are associated primarily with the Atlantic Multidecadal Oscillation (AMO), which is manifest in basin-wide sea surface temperature and sea level pressure fluctuations connected to large scale ocean circulation patterns. The AMO was in its warm phase during 1926-1970 and 1995-present, but in its cool phase during 1971-1994. It has its greatest impact on the number of major hurricanes (Category 3+), identified by Goldenberg *et al.* (2001) as associated with above normal SSTs and decreased vertical shear in the AMO warm phase (see also Bell and Chelliah, 2006; Klotzbach *et al.*, 2018).

Klotzbach *et al.* (2018) conducted a comprehensive evaluation of the landfalling hurricane data for the Continental U.S. since 1900. Figure 6.2.3 updates their analysis through 2024. While the largest numbers of landfalling hurricanes are from 2004, 2005 and 2020, there is no statistically significant trend since 1920. Figure 6.2.3 also shows the time series for major hurricane landfalls (Category 3-5). The largest year in the record is 2005, with 4 major hurricane landfalls. However, following 2005 there were no major hurricanes striking the U.S through 2016, the longest such period since 1920.

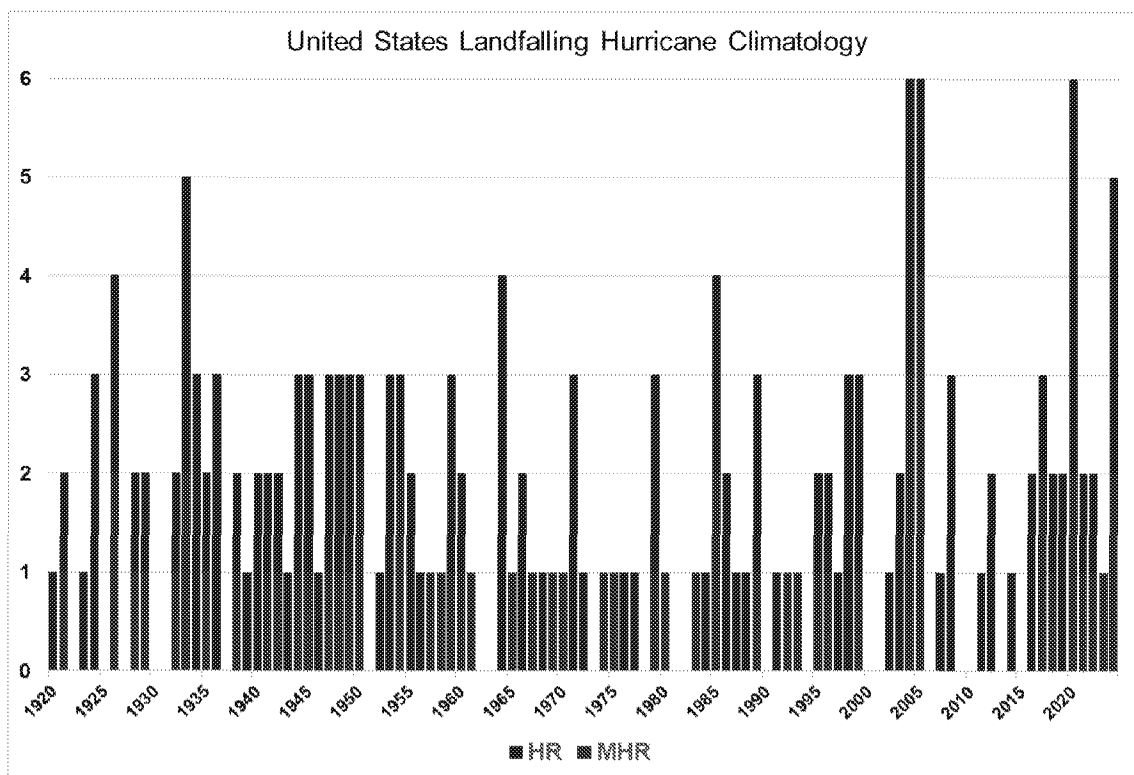


Figure 6.2.3: U.S. landfalling frequency of hurricanes (HR) and major hurricanes (MHR) since 1920. Source NOAA HRD(a) (2024)

Figure 6.2.3 shows substantial interannual to multidecadal variability in U.S. landfall activity. Klotzbach *et al.* (2018) examined how the landfall counts vary with ENSO (El Niño versus La Niña) and the warm versus cold phases of the Atlantic Multidecadal Oscillation (AMO).

Villarini *et al.* (2012) provide an analysis of U.S. hurricane landfalls back to 1878. While it is possible that some landfalls were missed in the late 19th century owing to sparsely populated regions on the Gulf Coast, it is remarkable that the highest year in the entire record is 1886, with 7 hurricane landfalls, when human influences on the climate were much smaller than they are today.

Table 6.2.1 shows the 10 strongest hurricanes (plus ties) to make U.S. landfall. Of the hurricanes that have made landfall with sustained winds greater than 150 mph, only one has occurred in the 21st century.

In summary, analyses of both global and regional variability and trends of hurricane activity provide the basis for detecting changes and understanding their causes. The relatively short historical record of hurricane activity, and the even shorter record from the satellite era, is not sufficient to assess whether recent hurricane activity is unusual relative to the background natural variability. Atlantic hurricane processes are influenced substantially by the natural modes of ocean circulation variability in the Atlantic, notably the Atlantic Multidecadal Oscillation. While it has long been hypothesized that a rising global sea surface temperature would cause an increase in hurricane intensity, identification of any significant trend in the hurricane data is hampered by a short data record and substantial natural variability.

Rank	Year	Landfall Wind (MPH)	Name
1	1935	185	"Labor Day"
2	1969	175	Camille
3	1992	165	Andrew
4	2018	160	Michael
5	1856	150	"Last Island"
5	1886	150	"Indianola"
5	1919	150	-----
5	1932	150	"Freeport"
5	2004	150	Charley
5	2020	150	Laura
5	2021	150	Ida
5	2022	150	Ian

Table 6.2.1 Strongest hurricanes to make landfall along the U.S. coast. Source (NOAA HRD(b), 2024)

6.3 Temperature extremes

The AR6 assessment focused on the period after 1950 and reported increasing trends in heatwave frequency and intensity. However, NCA4 noted that heatwave activity in the U.S. reached a peak in the 1930s (Figure 6.3.1).

AR6: It is *virtually certain* that hot extremes (including heatwaves) have become more frequent and more intense across most land regions since the 1950s, while cold extremes (including cold waves) have become less frequent and less severe (SPM, A3.1)

AR6: In North America, there is very robust evidence for a *very likely* increase in the intensity and frequency of hot extremes and decrease in the intensity and frequency of cold extremes for the whole continent, though there are substantial spatial and seasonal variations in the trends. Minimum temperatures display warming consistently across the continent, while there are more contrasting trends in the annual maximum daily temperatures in parts of the USA. (Chapter 11, p 1550)

NCA4: Changes in warm extremes are more nuanced than changes in cold extremes. For instance, the warmest daily temperature of the year **increased** in some parts of the West over the past century, but there were **decreases** in almost all locations east of the Rocky Mountains. In fact, all eastern regions experienced a net **decrease**, most notably the Midwest (about 2.2°F [1.2°C]) and the Southeast (roughly 1.5°F [0.8°C]). (pp. 190-191)

NCA4: Since the mid-1960s, there has been only a very slight increase in the warmest daily temperature of the year (amidst large interannual variability). Heat waves (6-day periods with a maximum temperature above the 90th percentile for 1961–1990) increased in frequency until the mid-1930s, became considerably less common through the mid-1960s, and increased in frequency again thereafter. As with warm daily temperatures, **heat wave magnitude reached a maximum in the 1930s.** (pp. 190-191)

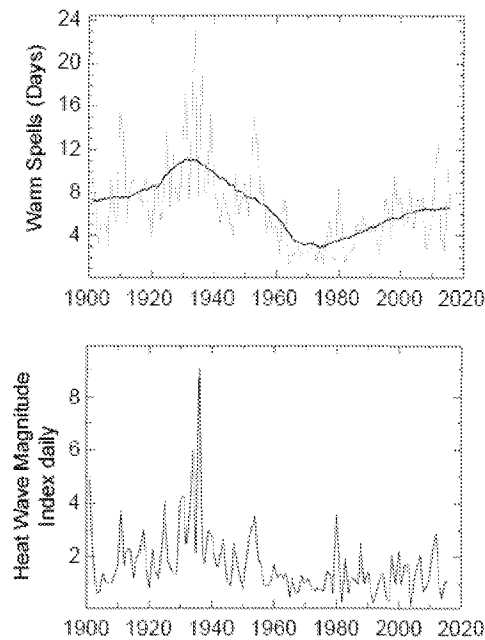


Figure 6.3.1: U.S. Heat waves since 1900. Source: NCA4 Figure 6.4

6.3.1 *Temperatures in the U.S. are becoming less extreme*

Daily maximum temperatures in the warm season (Tmax, May-Sep) and daily minimum temperatures in the cold season (Tmin, Dec-Mar) are available beginning in Dec 1898 (126 years). The dataset consists of 1,211 CONUS stations designated as United States Historical Climate Network or USHCN stations (see Figure 6.3.2; Quinlan *et al.* 1987, Karl *et al.* 1990). These stations were selected by NOAA as having the fewest problematic issues with gaps, station moves, and instrument changes. Where gaps still exist, nearby stations (bias-corrected) were merged so that the median volume of data available for a station is 98%. Although there are certainly errors in the dataset, including unresolved spurious warming due to UHI effects that especially bias Tmin records, this data set is sufficiently accurate for assessing trends in Tmax heat extremes.

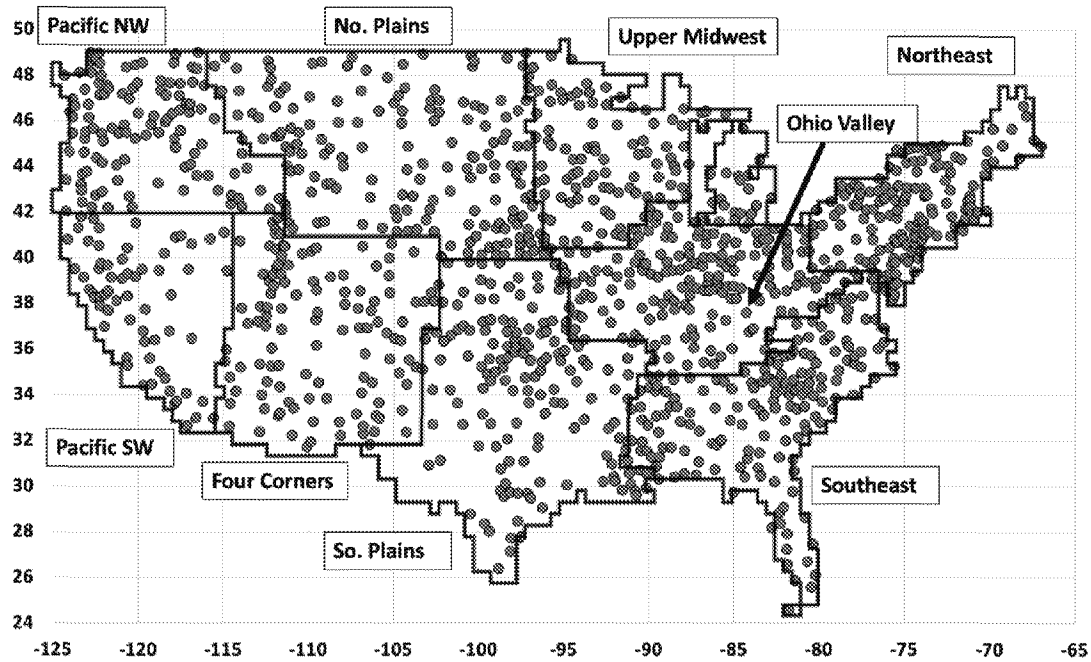


Figure 6.3.2 Locations of USHCN temperature stations. Source: USHCN.

We begin with the question of whether the occurrence of daily record high or low temperatures has changed since Dec 1898. Each warm-season has 153 days (1 May to 30 Sep) and each cold-season has 122 days (1 Dec to 31 Mar). For each station and day, we calculated the year in which the record highest (lowest) temperature occurred. With 126 years of observations, if there were no temperature trends over time, the expected number of records for Tmax would be 1.21 ($=153/126$) per station per year and for Tmin 0.96 ($=122/126$) per station per year.

Figure 6.3.3 shows the observed distribution in time of the occurrence of these extreme events. There is a common feature in many metrics of warm-season extremes in the CONUS - the exceptional heat of the 1920s and especially the 1930s, peaking in 1936. On a per-station average, 60 percent of the Tmax records and 59 percent of the Tmin records occurred in the first half of the period (1899-1961).

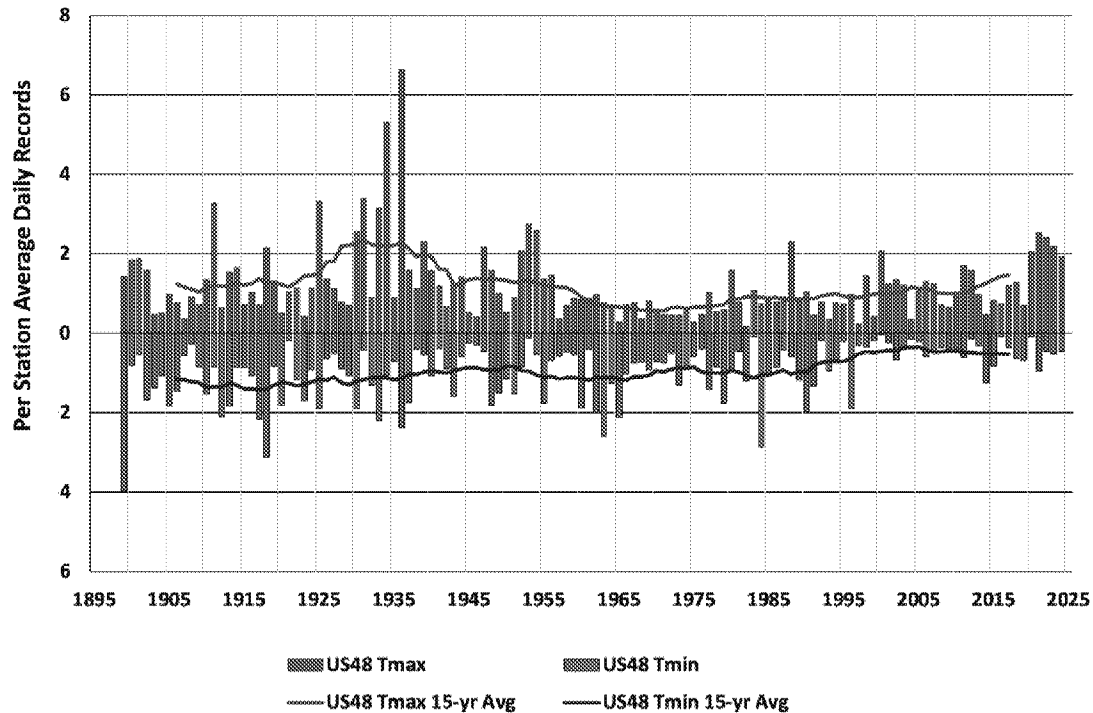


Figure 6.3.3 Number of daily record High (red) and Low (blue) temperatures for warm and cold seasons in the CONUS. The lines represent the 15-year running, centered average. Source: 1,211 USHCN stations supplemented as needed to achieve a minimum of 92 percent of observations in the 126-year period since Dec 1898. US48: contiguous U.S. states. Tmax: maximum temperature. Tmin: minimum temperature.

On the cold side, the Valentine’s Day Arctic outbreak in Feb 1899 stands as the most extensive cold extreme experienced by CONUS, with 1917 in 2nd place. The frequency of cold records has declined, especially over the last quarter of the period in which only 13 percent of the extreme cold events were measured. In contrast, 25 percent of the extreme Tmax records were achieved in the last quarter, in accordance with statistical expectations. These general features have been noted in past assessments (see above, IPCC AR6, NCA4). Combining the two histories, the overall reduction in numbers of both cold and hot extremes over the past century indicates a climate less prone to extremes.

This pattern is also shown in Figure 6.3.4. For each station and for each year the hottest warm season and coldest cold-season temperatures were calculated. Then the differences between these were computed by station and geographically-averaged over all stations, thus yielding an annual measure of the expected range of local extreme temperatures for each year. Figure 6.3.4 shows the 15-year trailing average of this measure, which has clearly declined over the past century.

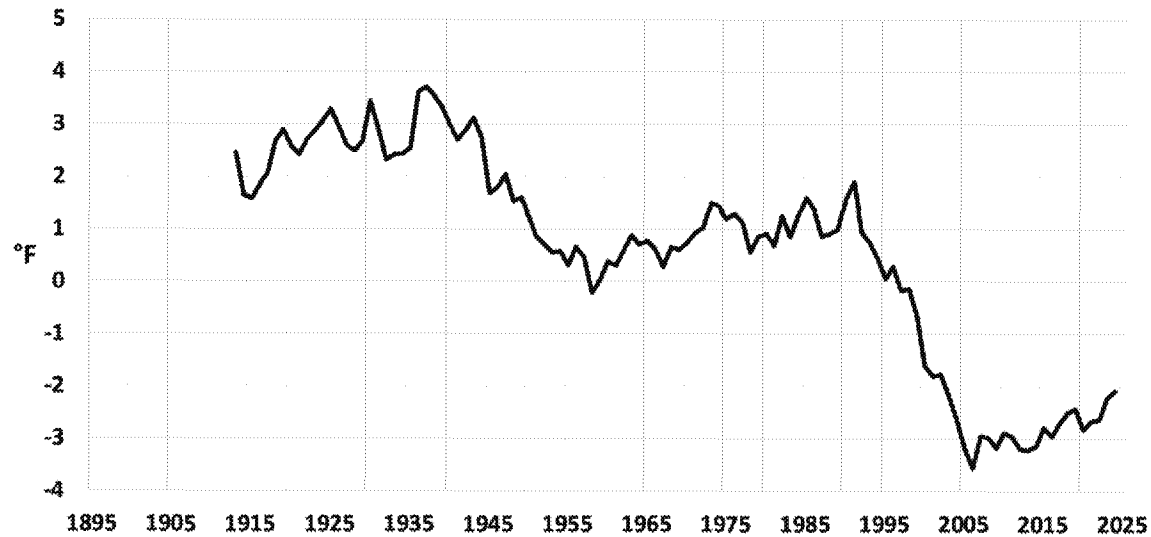


Figure 6.3.4. 15-year trailing average of the difference each year of each station's hottest warm-season Tmax and coldest cold-season Tmin relative to the long-term average. Source: Author analysis of USHCN data.

The average difference for each station between the hottest summer Tmax and coldest winter Tmin has declined by about 5°F in the past 126 years. The decline is due mostly to warmer winter Tmin, but a decline in summer Tmax is also a factor. The rise in Tmin has been strongly related to the growing presence of manufactured surfaces around the weather stations over the last 100+ years (the so-called urban heat island effect; Section 3.3 and Spencer *et al.* 2025).

In summary, while temperature extremes are regularly experienced in the U.S. and attract a great deal of media attention, long term records show the U.S. climate has become less extreme over time (milder) when measured by the range between warm season maxima and cold season minima.

6.3.2 Exceedances of a heat threshold

Under the heading of “The Risk of Temperature Extremes is Changing”, the most recent U.S. National Climate Assessment report (NCA5) notes the increase in a threshold metric of number of days at or above 95°F, stating,

The western U.S. has been particularly affected by extreme heat since the 1980s ..., experiencing a larger increase in days over 95°F, as would be expected given the greater warming in that region relative to the eastern US. Several major heatwaves have affected the U.S. since 2018, including a record-shattering event in the Pacific Northwest in 2021.

Are the occurrences of 95°F days changing? In a climate as varied as that of CONUS, threshold statistics can be misleading. A region with many stations that have near 95°F Tmax average temperatures in the summer might see large swings in the metric when only small changes in average temperature occur. Elsewhere with stations that either rarely or virtually always achieve 95°F Tmax temperatures, a small change will not have much impact on the results.

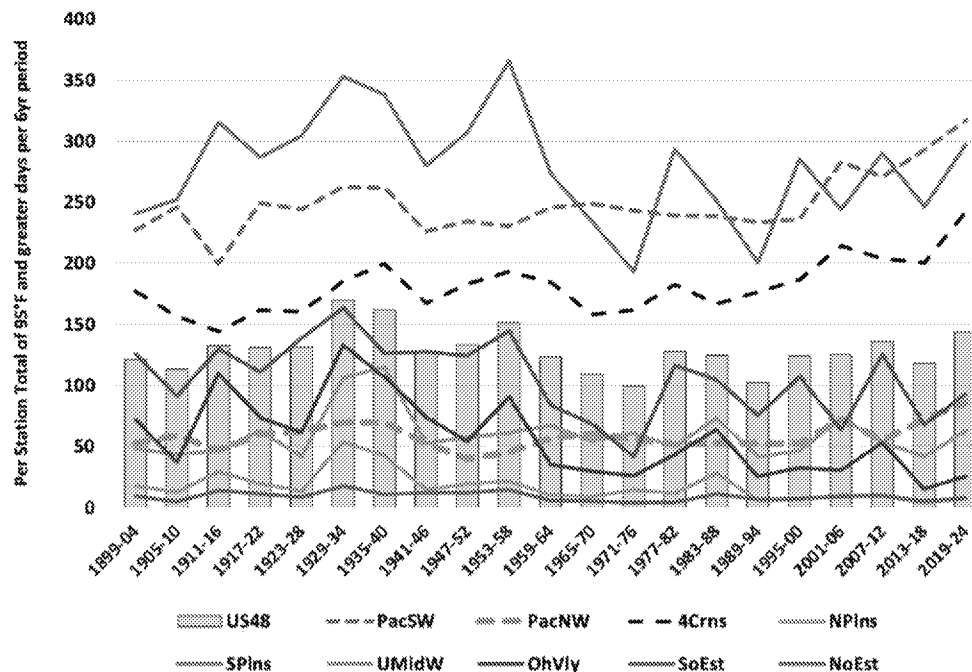


Figure 6.3.5. Total number of days $\geq 95^{\circ}\text{F}$ in 6-yr periods, U.S. 48 (bars) and regions (lines). A 6-year period is used as this evenly divides the 126-year record. US48: contiguous U.S. states. See Figure 6.3.2 for region names. Source: Author analysis of USHCN data.

In the past 126 years, the average CONUS station experienced 129 days exceeding 95°C per 6-yr period, but the regional values range from 278 in the Southern Plains to 9 in the Northeast. Thus, such threshold analyses must be interpreted with caution. Figure 6.3.5 shows that only three of the nine regions, all in the West, have experienced upward trends in the number of 95°F or hotter days (dashed lines). The CONUS as a whole has not, and the other six regions have experienced declines.

The Pacific NW heatwave of 2021 referenced in the NCA5 quote will be examined more closely in Section 8.6.1. The evidence indicates that it was a single, unprecedented event in the record, not part of a pattern of increasing extreme heat. For example, the 5-day average tropospheric grid-point temperature anomaly over the Pacific NW during that event was $+10.8^{\circ}\text{C}$, the most extreme Northern Hemisphere grid point summer anomaly in the 46 years from over 4 million grid values. In contrast, the global temperature anomaly during that time was virtually zero ($+0.03^{\circ}\text{C}$, Mass *et al.* 2024).

The NCA5 heading “The Risk of Temperature Extremes is Changing” suggests pervasive positive trends are now being observed in threshold temperatures, but that is not evident in the figures above.

6.3.3 Heatwaves

Heatwaves (consecutive days that exceed an extreme threshold) have a greater societal impact than a single daily record temperature. We measure “Heatwave Days” here as the count of all days in May-Sep each year that exceed the 90th percentile for that day and that lie within a period of at least six consecutive days. This is equivalent to the method used in NCA4, except that the reference period here is the entire record 1899-2024 while NCA4 truncated the reference period to 1961-1990, which was a cool interval. That truncation boosts positive results (days exceeding the 90th percentile) in years warmer than the reference period, especially starting in 1960 and moving to the present (see Figure 6.3.6 and discussion below).

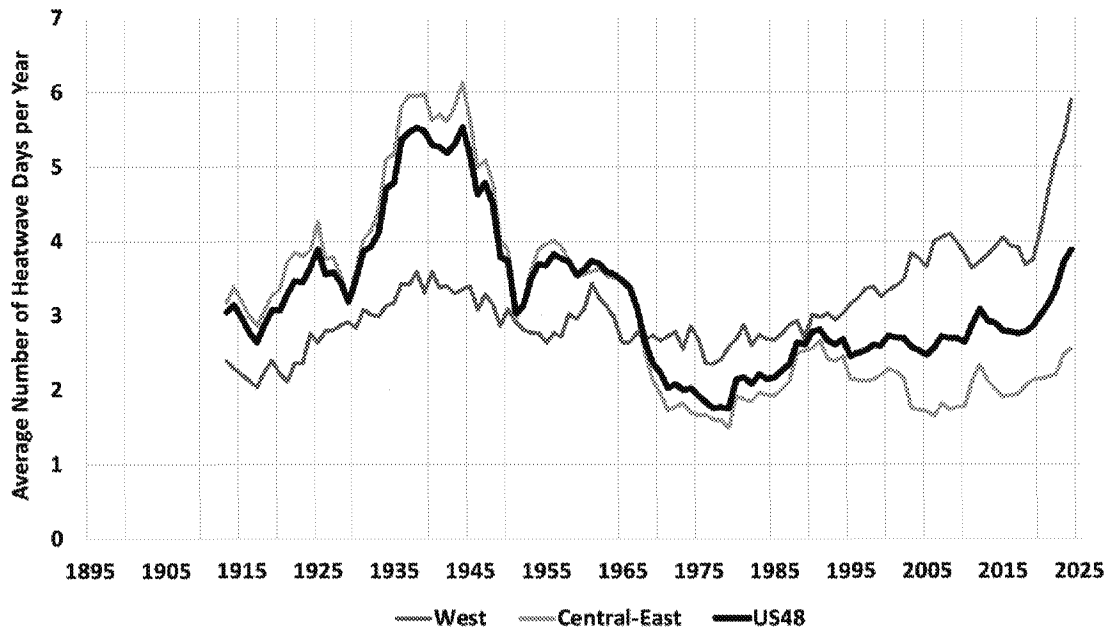


Figure 6.3.6 15-year trailing average of number of heatwave days per year per station in the CONUS (black line) and two regions: West (red), Central-east (green).

Figure 6.3.6 indicates that there are regional variations in heatwave activity. The excessive heat of the first half of the 20th century occurred primarily in the eastern two-thirds of the nation, while the West has seen a recent increase of heatwave days (NCA5). This indicates that the background warm season circulation favored heatwaves in the eastern portions of the country in the first half of the 20th century, but in the 21st century the patterns have favored heatwaves in the West. For CONUS as a whole, heatwaves are no more common today than they were a century ago, consistent with the upper panel of our Figure 6.4.1 taken from NCA4.

This metric varies significantly with region. The four northern regions (Pacific NW, Northern Plains, Upper Midwest, and Northeast) on average experience 15 to 27 heatwave days per 15-year period. In contrast, the five southern regions (Pacific Southwest, 4-Corners, Southern Plains, Ohio Valley and Southeast) see 37 to 54 such days, essentially twice as many. This suggests the summer circulation pattern is more prone to stationary events in the southern regions while transient systems in the northern regions are more common and thus cut short these potentially longer events.

The analysis of heatwaves is an example of why it is important to consider complete datasets and appropriate metrics. The NCA5 directs readers to the website [[HYPERLINK "https://www.globalchange.gov/indicators/heat-waves"](https://www.globalchange.gov/indicators/heat-waves)] (USGCRP 2023) to view a figure showing the number of urban heatwaves by decade from the 1960s, which we reproduce as Figure 6.3.7.

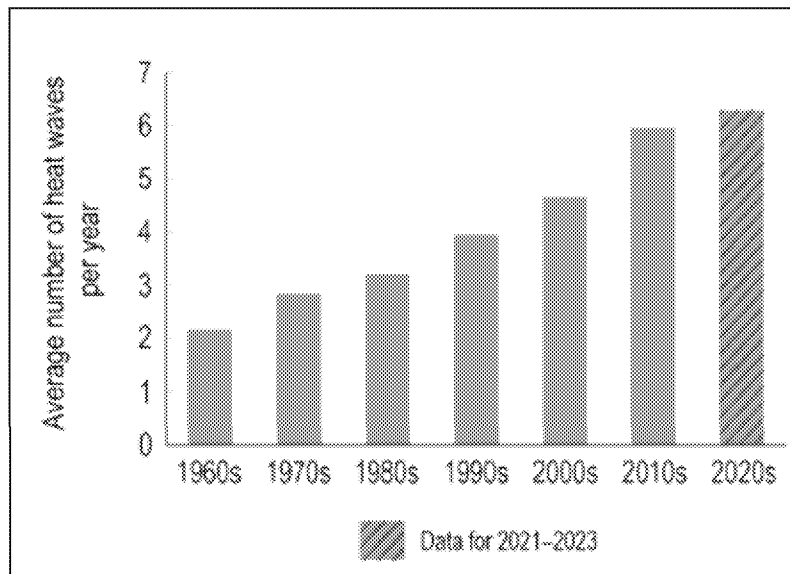


Figure 6.3.7: Average number of urban heatwaves per year for 50 large U.S. metropolitan areas, a misleading metric for reasons explained in the text. From [[HYPERLINK "https://www.globalchange.gov/indicators/heat-waves"](https://www.globalchange.gov/indicators/heat-waves)] (accessed May 22, 2025).

The figure shows a monotonic increase in each decade from two occurrences per year in the 1960s to six in the 2020s. The definition of a heatwave used is an unusual but practical measure of human discomfort - a period of at least 2 consecutive days when the *minimum* apparent temperature (combination of temperature and humidity) exceeds the 85th percentile. Note too, the dataset is limited to the 50 largest U.S. cities.

Given the unusual heatwave definition and urban focus, these increasing values since 1960 presented in USGCRP (2023) are not informative about long term trends or the influence of GHG emissions for at least two reasons. First, as shown in Figures 6.3.1 and 6.3.6, the 1960s was the coldest decade and 1970s the second coldest decade since the 1910's, so this starting date preconditions the time series to show increases. Second, post-1960 urbanization in these cities is a major factor in the rise of T_{min} relative to co-located T_{max} and relative to T_{min} at nearby rural stations. This does not dismiss the real rise in nighttime temperatures in major U. S. cities and the societal impacts associated with these changes. However, we note for a variety of reasons that summer T_{max} (especially in rural areas) is a better metric for detecting changes in heatwaves influenced by changes in the background climate due, for example, to increasing GHGs (Christy *et al.* 2009). For CONUS as a whole, the evidence in this section suggests GHG emissions have had little-to-no effect on heatwaves against the background of urbanization and natural climate variability.

6.4 Extreme precipitation

AR6 assessed that an increase in heavy precipitation has been observed in data starting in the 1950s.

AR6: The frequency and intensity of **heavy precipitation events have increased** since the 1950s over most land area for which observational data are sufficient for trend analysis (high confidence). (SPM A3.2)

AR6: In North America, there is robust evidence that the **magnitude and intensity of extreme precipitation has *very likely* increased** since the 1950s. Both [one-day maxima] and [5-day maxima] have significantly increased in North America during 1950-2018. (Chapter 11, p. 1560)

The U.S. National Climate Assessments (NCA4, NCA5) have highlighted an increase in the occurrence of the heaviest precipitation events (defined in different ways) primarily in the eastern half of CONUS, especially the Northeast, when starting the analysis in either 1901 or 1958. Interestingly, the regional variations indicate that the largest increases in extreme precipitation events are in the Northeast and the smallest in the West, a pattern counter to the changes in temperature extremes (Figure 6.3.6).

McKittrick and Christy (2019) examined long-term and consistent station observations of extreme daily precipitation to test some of these NCA claims for the Southeast and West Coast using a trend model with a non-parametric variance estimator robust to the complex autocorrelation properties of precipitation data. When the time series were extended back in time (as far as 1872 in some cases) or started later (1978), there were no significant trends for either region.

These findings have been updated for this report (McKittrick and Christy 2025) with similarly constructed observations from 29 stations on the CONUS Pacific Coast (1893ff from San Diego CA to Blaine WA) and 24 stations in the humid Southeast (1872ff from Austin TX to Washington DC), also adding 27 stations in the Northeast (1888ff from Buffalo NY to Eastport ME). The locations are shown in Figure 6.4.1. The stations were selected based on availability of long-term high-quality records. The regions are each associated with important features of extreme precipitation behavior: the Pacific coast is associated with landfalling atmospheric rivers for which AR6 cites evidence of increasing activity since 1948 with further increases expected as the world warms (AR6 8.3.2.8.2); the NCA report indicates that the Northeast has experienced the greatest increase in extreme events, and the Southeast is also a place noted in the NCAs as having increased extreme events.

The results of applying the analysis of McKittrick and Christy (2019) were as follows for each region, then followed by further explanation.

Pacific Coast heavy rainfall events

- The average precipitation trend is statistically significant (downwards) in Astoria OR; insignificant elsewhere.
- The trend in rainfall variance is positive and significant in Big Sur CA; insignificant elsewhere.
- The trend in daily maximum precipitation is positive and significant in Aberdeen WA and Big Sur CA and negative and significant in Newport OR (insignificant elsewhere).
- Averaged over all stations in the region, none of these three trend parameters is statistically significant.

The Pacific Coast receives considerable precipitation from Atmospheric River (AR) events which often last more than a day or two (e.g., Gershunov *et al.* 2017, Pan *et al.* 2024). The worst series of such events in recent history was the so-called ARkStorm that occurred during December 1861 and January 1862; it dumped nearly 10 feet of rain in parts of California and submerged the entire Central Valley for weeks under as much as 15 feet of water (Brewer 1930, Null and Hulbert 2007). Additionally, Paleoclimate research has found six megastorms more severe than 1861–1862 in California during the last 1800 years, occurring at intervals of 300 years or so (Porter *et al.* 2011).



Figure 6.4.1: Locations of precipitation monitoring stations used in this report. Orange: Pacific coast. Blue: Northeast. Green: Southeast. Data from McKittrick and Christy (2025).

We examine occurrences of 5-day deluges as follows. Taking the Pacific coast as an example, a 130-year span contains 26 5-year intervals. At each location we computed the 5-day precipitation totals throughout the year and selected the 26 highest values across the sample. A single year might have more than one of the 26 heaviest. Each of those can be thought of as 1-in-5yr events. If there are no trends in precipitation, then the total number of these events across all stations should be evenly spread over the years. In Fig. 6.4.2 we show the distribution in time of these events for the Pacific coast. The deluges associated with the massive 1997-98 El Niño event are readily apparent. While erratic, as is typical of such precipitation metrics, there is no indication of a tendency to become more frequent over time.

Southeast heavy rainfall events

- The trend in average precipitation is positive and statistically significant in Mobile AL and Quitman GA but insignificant elsewhere.
- The trend in rainfall variance is positive and significant in Mobile AL but insignificant elsewhere.
- The trend in daily maximum precipitation is positive and significant in Vicksburg MS and Norfolk VA but insignificant elsewhere.
- Averaged over all stations in the region none of these three trend parameters is statistically significant

Figure 6.4.3 is analogous to Figure 6.4.2 for the last 150 years in the Southeast humid zone. The temporal pattern of 5-day totals of the 1-in-5yr heavy events is generally unremarkable, though a cluster of higher values appears in 1995 to 2019. The increase in those years is due largely to the 4 northeastern-most stations of Wilmington NC, Weldon NC, Washington DC, and Norfolk VA. This confirms the pattern indicated in NCA4 and NCA5 -- an increasing frequency of heavy events due to a temporal clustering of tropical storms from eastern NC to Maine discussed below. Otherwise, the remaining 20 stations show an unremarkable temporal distribution of heavy events.

BOX: Perils of short data records

San Francisco provides a good case study of the limitation of using short historical samples to characterize natural variability of extreme events. Suppose we use a 130-year sample of daily San Francisco precipitation from 1895 to 2024 and we look for 3-day, 5-day, 14-day and 30-day rainfall records. The results are as shown in Table 7.2.

Event	Record (inches)	Year
3-day	6.94	2023
5-day	8.55	2023
14-day	12.62	2023
30-day	18.93	1998

Table 6.2: Extreme rainfall records, San Francisco, 1895-2024.

The records all cluster in the more recent years. 2023 appears to be an exceptional year and since it is near the end of the sample, it might suggest that the climate has shifted into a more hazardous state, perhaps because of human influences. [Note “2023” indicates the event occurred in the water-year of Aug 2022 to Jul 2023.]

But the picture is very different if we use a sample that begins 45 years earlier, in 1850. Table 6.3 shows that the record-setting events all happened in the 1860s. Furthermore 2023 is now not even in 2nd place but falls to 3rd or 4th place. And comparing the records of the two charts shows the extreme precipitation events in 1862 and 1867 involved considerably more rainfall than the 1998 and 2023 events, with 14- and 30-day totals about 50 percent higher.

Event	Record (inches)	Year	Rank of 130-yr extreme listed in Table 6.2
3-day	8.85	1867	3
5-day	9.80	1867	3
14-day	19.05	1862	4
30-day	28.25	1862	2

Table 6.3: Extreme rainfall records, San Francisco, 1850-2024.

The range of natural variability is made even more remarkable when paleoclimate evidence is examined. Porter *et al.* (2011) discovered that in the past 1,800 years at least six megastorms were more intense than the devastating 1861-62 ARkStorm that struck the region. Evidently such extreme events, “unprecedented” in our 1895-2024 sample, have impacted the region about every 300 years, though not since 1895.

This example illustrates the limitations of using relatively short climate periods (~130 years) to assess the character and range of natural variability in general and of extreme events in particular. Accurate representation of the full range of natural variability is necessary for any attribution analyses (Section 8.6). Infrastructure planners, emergency management institutions and attribution scientists would understand the significant mischaracterization of the magnitude of a future extreme if based only on the last 130 years. In this case, a single time-sample of 130 years provides an underestimation of the extreme value by up to 50 percent determined when adding only 45 more years of observations. Compared to millennial-scale paleoclimate evidence, an even greater underestimation would occur. An important lesson is that the climate can deliver great surprises on its own, even without human influences.

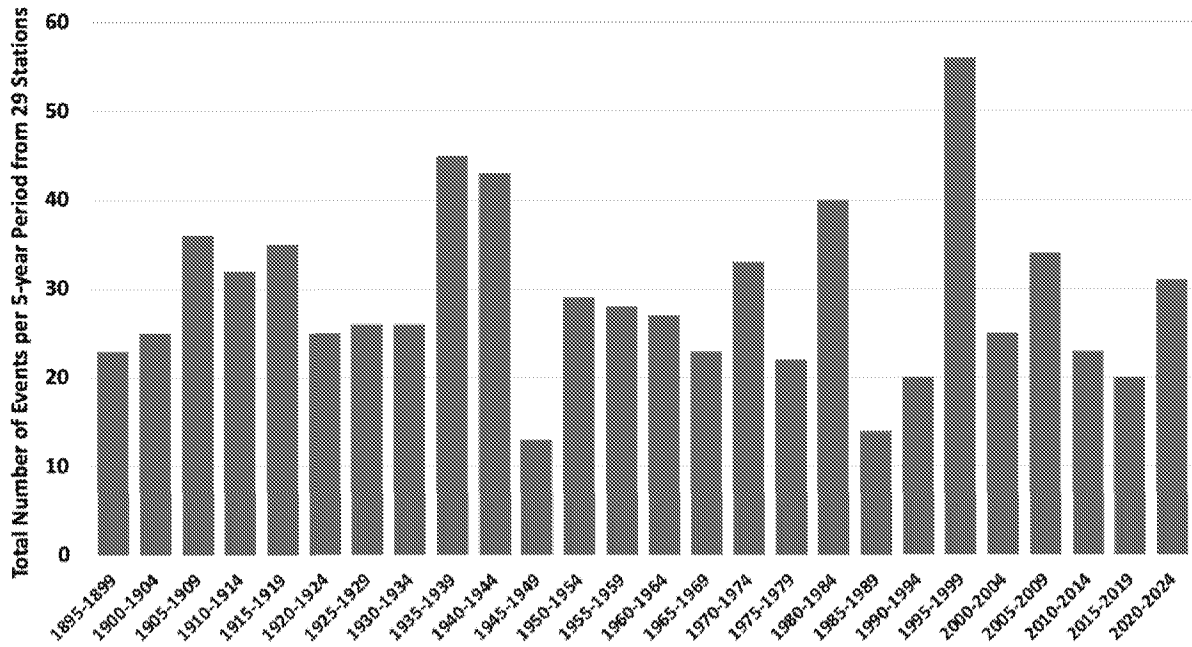


Figure 6.4.2. The time distribution by 5-year periods of the 26 heaviest (1-in-5 yr) occurrences for 29 stations on the Pacific coast.

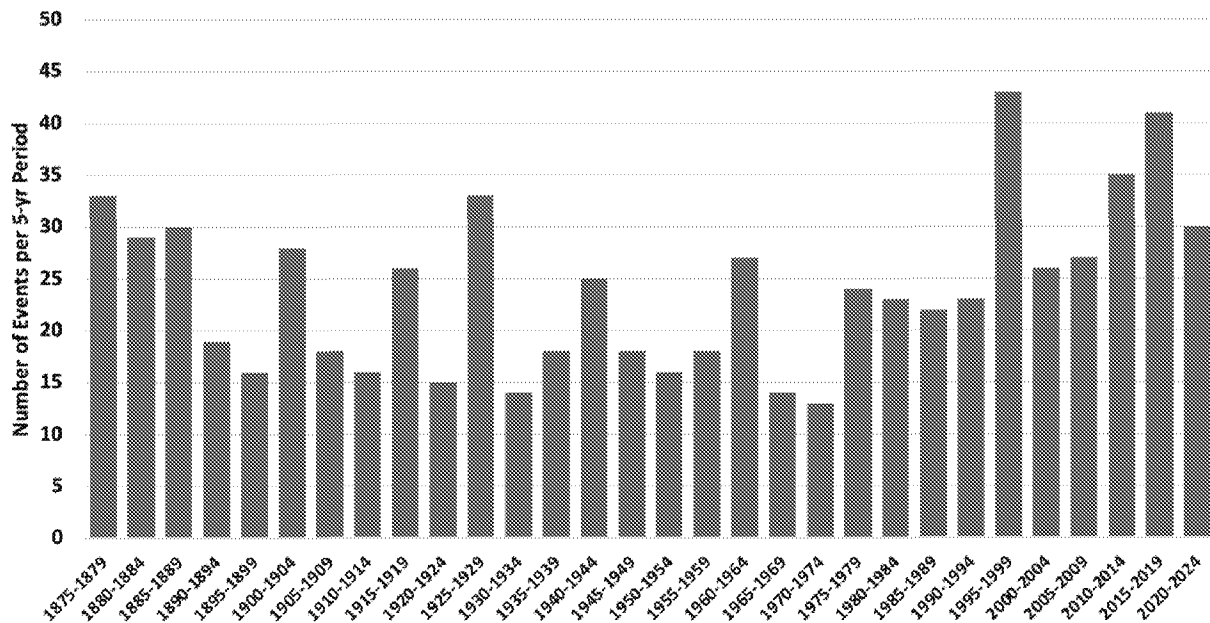


Figure 6.4.3. As in Fig. 6.4.2 but for the heaviest 30 (1-in-5yr) events for 24 stations in the humid Southeast from Austin TX to Washington DC in 5-year bins for 1875-2024.

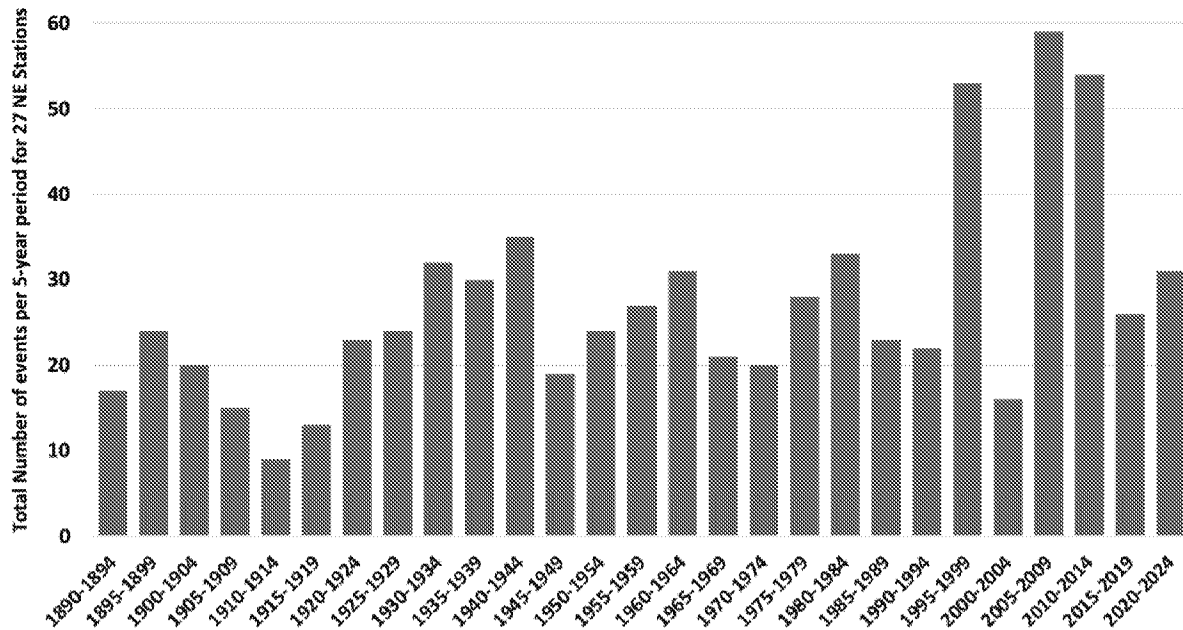


Figure 6.4.4. As in Fig. 6.4.2 but for the heaviest 27 (1-in-5yr) 3-day precipitation events for 27 Northeast stations from NY to ME, including Montreal.

Northeast heavy rainfall events

- The trend in average precipitation is positive and statistically significant in 12 of 27 locations and also in the regional average.
- The trend in rainfall variance is positive and significant in Portland ME, Albany NY, Buffalo NY and Eastport ME but insignificant elsewhere.
- The trend in daily maximum precipitation is positive and significant in Portland ME, Gardiner ME and Eastport ME but insignificant elsewhere.
- When averaged over all stations in the region, there is no statistically significant trend in either the precipitation variance or maximum

Fig. 6.4.4 is analogous to Figure 6.4.2 for the last 135 years in 27 stations in the Northeast (including Montreal Canada). We use 3-day totals here as this produced the largest temporal variations in time. In this region, 77 percent of events occur during June to October and are dominated by incursions of hurricanes, tropical storms, or tropical storms that transition to extratropical systems. According to NCA4 and NCA5 this region experienced the largest increases in extreme events, so it merits a closer examination.

There is a noticeable clustering of extreme events from 1995 to 2014. Howarth *et al.* (2019) examined a similar region as in Fig. 6.4.4 that includes PA and NJ and reported significant differences in various precipitation extremes between two 18-year periods, 1979-96 and 1997-2014. That included a 317 percent increase in 24-hr events exceeding 6 inches, while we find a 58 percent increase over the same years. However, Figure 6.4.4 shows that frequency drops sharply after 2014, returning to the long-term average in the subsequent 5-year intervals, again illustrating the perils of drawing conclusions from short-term trends in highly variable metrics.

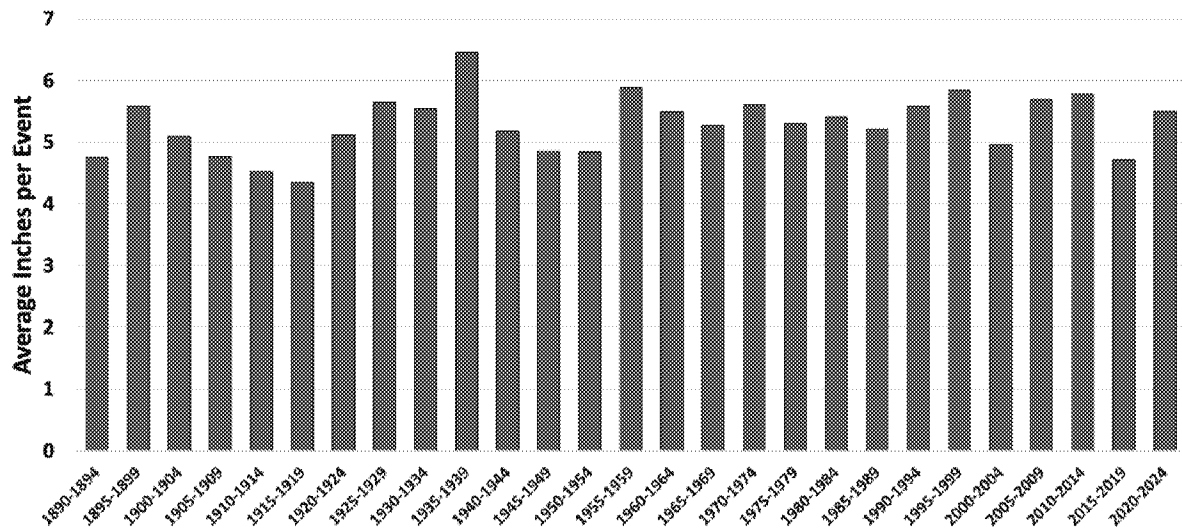


Figure 6.4.5. Average amount of precipitation falling in the 1-in-5yr events for the NE stations.

The high percentage increase in the Howarth *et al.* sample is associated with small numbers at relatively few locations: there were only 6 in the first period and 25 in the second across 58 stations. Most of the stations did not record such an event. If there is a region-wide increase in heavy events, it should be seen in the average across all stations. Figure 6.4.5 shows the average precipitation in the 1-in-5 yr events across the NE. The trend is only +0.04 inches/decade. The highest amount, 1935-39, includes the Great New England Hurricane of 1938, one of the rare major (Category 3 or higher) hurricanes to strike the region. The results in Figs. 6.4.4 and 6.4.5 thus suggest that though there was a surge in the number of events at a few locations during the 1995 to 2014 interval, there was no regional pattern and the change did not persist beyond 2014.

Jong *et al.* (2024) document the increase of tropical influences on precipitation events in the Northeast since 1959 and concluded “The autumn extreme precipitation trend over the Northeast U.S. is primarily attributed to tropical cyclone-related events since the 1990s.” The question then becomes: “Was the temporal clustering of tropical systems in 1997-2014 which affected the Northeast a response to increasing GHGs?” Jong *et al.* examined CMIP-6 model output which suggests that there will be fewer such systems in the 21st century but that the intensity of the rainfall events might increase. This conjecture is not seen in Fig. 6.4.5 where the amount-per-event has remained steady over the 135-year period.

There is some evidence to indicate the heaviest rainfall events might be redistributed due to the impact of urban infrastructure on the local weather (e.g., Pielke Sr. *et al.* 2011, Zhang *et al.* 2018, Yang *et al.* 2024). Yang *et al.* state “Cities that experience compact development tend to witness larger increases in extreme rainfall frequency over downtown than their rural surroundings, while the anomalies in extreme rainfall frequency diminish for cities with dispersed development.” While this is an important insight to consider, the effect on the specific stations used in this analysis is unknown, or at least not detectable in Fig. 6.4.5.

In summary, some U.S. regions show short-duration increases in extreme precipitation events, consistent with natural variability. But analysis of long term, nationwide historical records that considers the autocorrelation properties of precipitation data does not support the claim that extreme short-duration rainfall events are becoming more frequent or intense.

6.5 Tornadoes

AR6 assesses tornado trends in the U.S. as follows:

[O]bservational trends in tornadoes, hail, and lightning associated with severe convective storms are not robustly detected due to insufficient coverage of the long-term observations. There is *medium confidence* that the mean annual number of tornadoes in the USA has remained relatively constant. (Chapter 11, section 11.7.3, p. 1594)

The monitoring of weak tornadoes has changed over time. The growth of rural populations and the increasing ability to take video with hand-held devices has led to more frequent reports of weak tornadoes that produce minimal damage. In contrast, strong to violent tornadoes have been observed more consistently over time. Note that tornado strength is measured by the damage it produces, not by the visual appearance of the funnel. Limited real-time observational capabilities in earlier decades did not prevent identification because strong to violent tornadoes leave much more damage which will be assessed later even if the tornado itself was not observed. Since statistics began in 1950, there has been a substantial decrease (by about 50%) in the number of strong to violent tornadoes as shown in Fig. 6.5.1a.

To summarize, there is a noticeable downward trend in the number of severe tornadoes in the U.S. since 1950. After 1990 the number of weak tornadoes in the U.S. has remained roughly constant; data before that are incomplete due to limited monitoring.

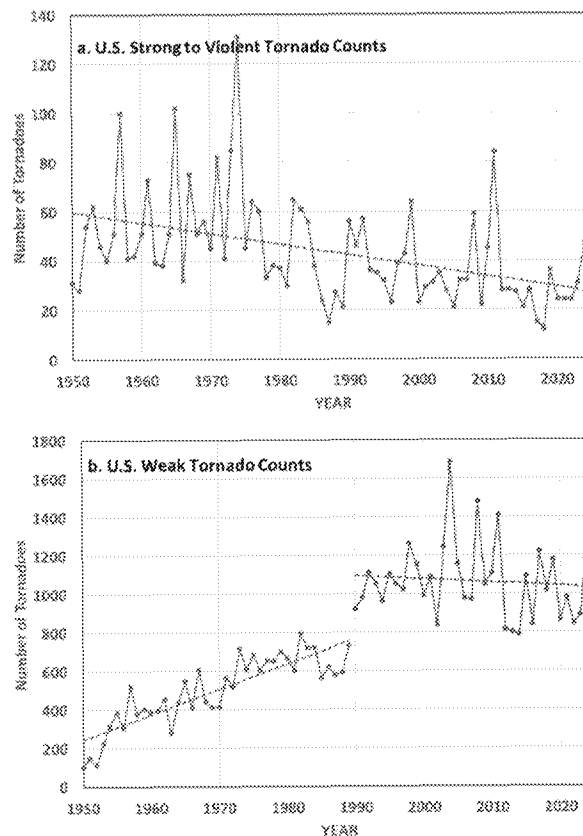


Figure 6.5.1. Annual U.S. tornado counts for (a) strong to violent tornadoes (EF3 to EF5), and (b) weak tornadoes (EF0 to EF2). Based upon NOAA Storms

Prediction Center data, available at https://www.spc.noaa.gov/wcm/data/1950-2024_actual_tornadoes.csv

6.6 Flooding

Changes in floods were assessed as follows:

AR6: The SREX **assessed low confidence for observed changes in the magnitude or frequency of floods** at the global scale. This assessment was confirmed by the AR5 report. The SR15 found increases in flood frequency and extreme streamflow in some regions, but decreases in other regions. . . [H]ydrological literature on observed flood changes is heterogeneous, focusing at regional and sub-regional basin scales, making it difficult to synthesise at the global and sometimes regional scales. (Chapter 11.5)

AR6: [T]he seasonality of floods has changed in cold regions where snowmelt dominates the flow regime in response to warming (*high confidence*). Confidence about peak flow trends over past decades on the global scale is *low*. (Chapter 11.5)

NCA4: Trends in extreme high values of streamflow are **mixed** across the United States. Analysis of 200 U.S. stream gauges indicates areas of both increasing and decreasing flooding magnitude **but does not provide robust evidence** that these trends are attributable to human influences (pp. 240-241)

The absence of detectable US-wide trends in flooding is consistent with the findings in Section 6.4 of absence of coherent changes in extreme precipitation.

6.7 Droughts

Assessments of drought trends were as follows.

AR6: Few AR6 regions show observed increases in meteorological drought (Section 11.9, p. 1575),

AR6: Increasing trends in agricultural and ecological droughts have been observed on all continents (*medium confidence*), but decreases only in one AR6 region (*medium confidence*). Increasing trends in hydrological droughts have been observed in a few AR6 regions. (Chapter 11 Summary)

NCA4: As a consequence of this increased precipitation, **drought statistics over the entire CONUS have declined**. (p. 233)

NCA4: Recent droughts and associated heat waves have reached record intensity in some regions of the United States; however, by geographical scale and duration, the Dust Bowl era of the **1930s remains the benchmark** drought and extreme heat event in the historical record (very high confidence). (p.231)

SREX: From a paleoclimate perspective, **recent droughts are not unprecedented**, with severe ‘megadroughts’ reported in the paleoclimatic record for Europe, North America, and Australia. (p. 170)

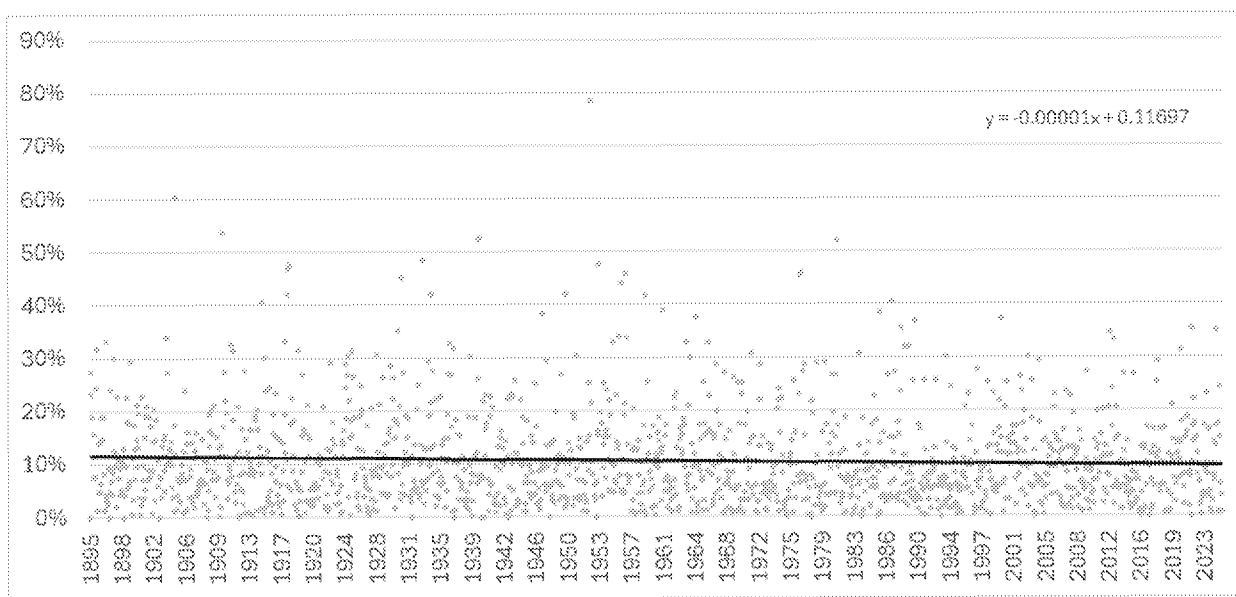


Figure 6.7.1: Monthly percent of US classified as “Very Dry” 1895—2025. Data source: NOAA [[HYPERLINK "https://www.ncei.noaa.gov/access/monitoring/uspa/wet-dry/0"](https://www.ncei.noaa.gov/access/monitoring/uspa/wet-dry/0)] least squares trendline added.

As shown in Figure 6.7.1, U.S. long-term data shows an insignificantly declining trend in extreme dryness (-0.001 percent per year)

Kogan *et al.* (2020) examines a 38-year high-resolution satellite-based drought measure and concludes that global drought has not intensified and is not connected to climate change: “it is possible to state firmly that global and main grain countries’ drought area and intensity trends have not been following global climate warming since 1980’s.”

In summary there is no evidence of increasing drought frequency or intensity in the U.S. or globally over recent decades.

6.8 Wildfires

The IPCC has not provided an assessment of wildfires. As shown in Figure 6.8.1, global wildfire activity as measured by European Space Agency shows a downward trend in the 21st century.

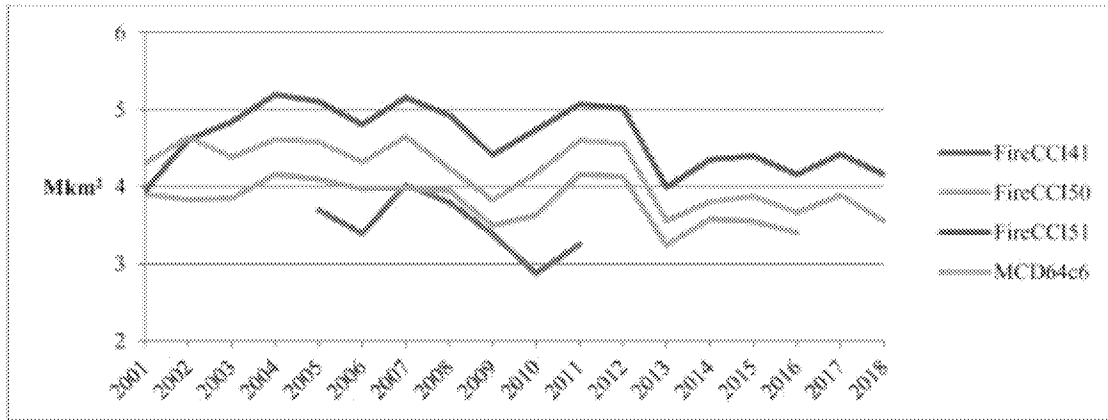


Figure 6.8.1: Global wildfire area 2001-2018. Source: From Lizundia-Loiola *et al.* (2021) Figure 12. The different coloured lines represent data products derived from different satellites and algorithms

Global data show that wildfire coverage is constant or declining on every continent (Samborska and Ritchie, 2024).

Active fire suppression since 1900 makes it difficult to establish a natural baseline for wildfire activity in the U.S. Paleoclimatic evidence indicates that past activity was much higher than today. Marlon *et al.* (2012) used sedimentary charcoal layers to reconstruct fire history of western U.S. for the past 1400 years and also fit a model to predict fire activity as a function of climatic conditions. Their results are summarized in Figure 6.8.2 below (from Figure 2 in their paper). There has been a growing wildfire deficit over the 20th century. In other words, however much fire was observed in the 20th century, it was less than what would have been observed in previous centuries based on the climatic conditions. Parks *et al.* (2025) likewise find that despite the recent increase in wildfire burn area in North America, a significant wildfire deficit remains relative to historical wildfire regimes.

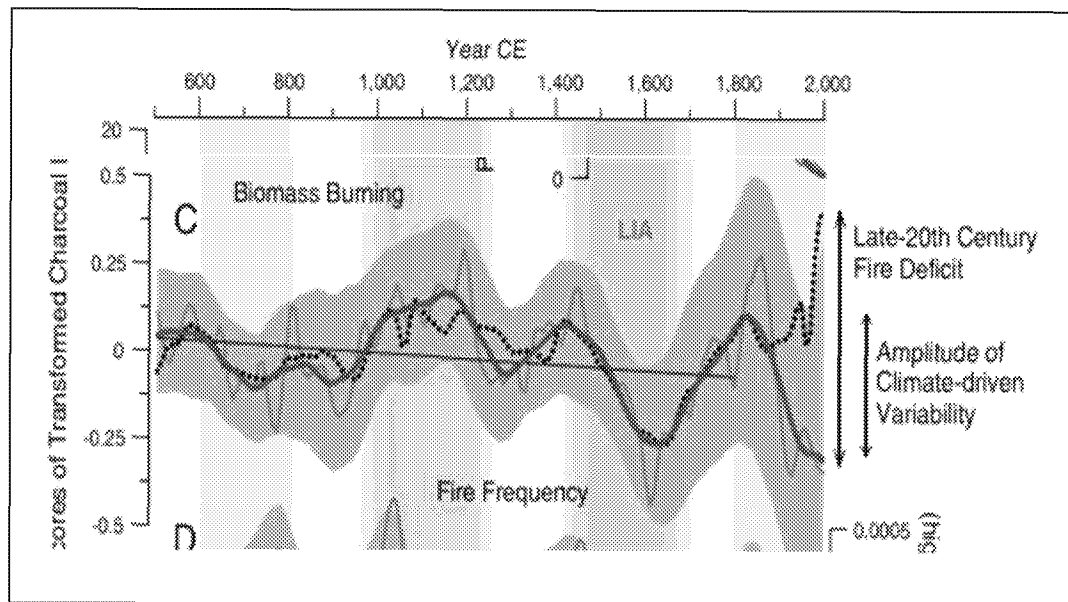


Figure 6.8.2 Fire frequency and fire deficit in the US. The red line shows the smoothed charcoal record and the black dotted line shows the predicted charcoal record based on climatic conditions. Source: Marlon *et al.* (2012) Figure 2.

U.S. data from the National Interagency Fire Centre (NIFC) from 1926 to 2023 are shown in Figure 6.8.3. The NIFC has removed the pre-1960 data from its current website on the grounds that measurement methods changed after 1960 making the comparison unreliable. Nonetheless just focusing on the post-1985 interval the number of fires is not increasing. The area burned did increase but only until about 2007.

Forest fires have always been a part of nature, and they can certainly create conditions that are inhospitable in the short term for all life, including humans. Science has confirmed the overall benefit and necessity of forest fires. While recent high-profile fires and seasons serve as a reminder of the potential destructive impact, the highest profile U.S. forest fire remains the 1910 Big Blowup fire in the U.S. west, which destroyed over three million acres and eliminated entire towns like Taft, MT (Apple, 2020). The 1910 fire reshaped the U.S. Forest Service (National Forest Foundation 2022) leading to a focus on fire suppression with a primary goal of defeating all forest fires (Forest History Society, 2022). This led to the “10 am rule” in 1935 requiring that all fires spotted on any day had to be controlled by 10 am the following day (National Forest Foundation, 2022).

While defeating all fires seemed a noble goal, questions began to arise as to whether this behavior “followed the science” (U.S. Forest Service, 2022). Over time the U.S. Forest Service has begun to rethink its goals, recognizing that new approaches such as prescribed burns, fuel elimination, and controlled wildfires are more appropriate (Sommer, 2016). Recent research is validating this approach and recognizing that more frequent smaller fires likely result in healthier forests, water ecosystems and biodiversity (Stephens *et al.*, 2021).

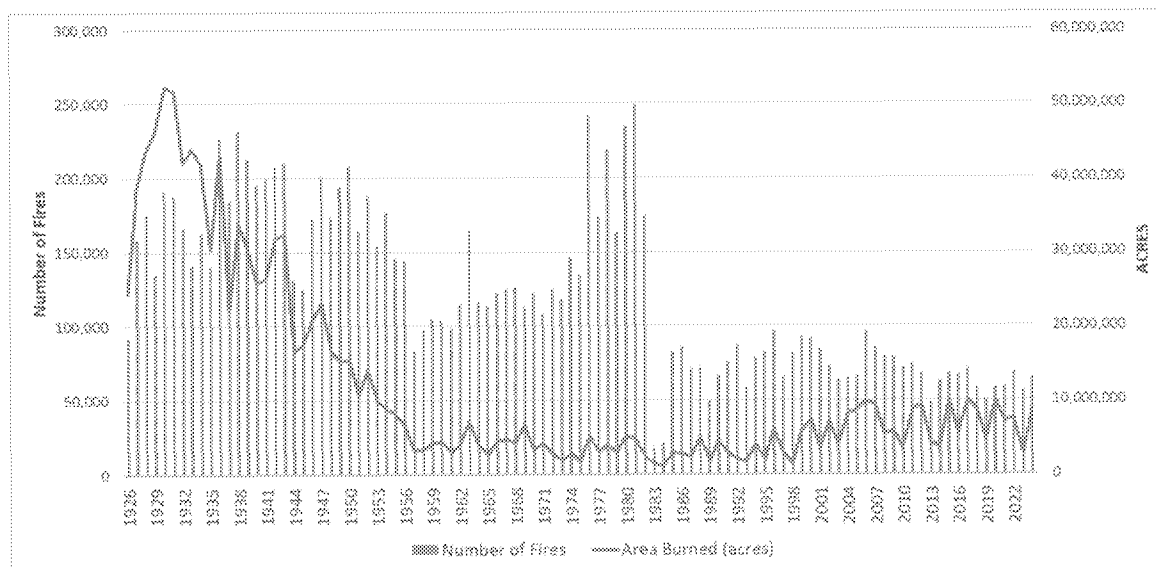


Figure 6.8.3: U.S. wildfires 1926 to 2023. Source: Post-2018: National InterAgency Fire Center data [[HYPERLINK "https://www.nifc.gov/fire-information/statistics/wildfires"](https://www.nifc.gov/fire-information/statistics/wildfires)]. Pre-2017 webarchive.org (n.d.).

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7 CHANGES IN SEA LEVEL

Chapter Summary

Since 1900, global average sea level has risen by about 8 inches. Sea level change along U.S. coasts is highly variable, associated with local variations in processes that contribute to sinking and also with ocean circulation patterns. The largest sea level increases along U.S. coasts are Galveston, New Orleans, and the Chesapeake Bay regions - each of these locations is associated with substantial local land sinking (subsidence) unrelated to climate change.

Extreme projections of global sea level rise are associated with an implausible extreme emissions scenario and inclusion of poorly understood processes associated with hypothetical ice sheet instabilities. In evaluating the IPCC AR6 projections to 2050 (with reference to the baseline period 1995–2014), almost half of the interval has elapsed by 2025, with sea level rising at a lower rate than predicted. U.S. tide gauge measurements reveal no obvious acceleration beyond the historical average rate of sea level rise.

7.1 Global sea level rise

Global sea level rise is arguably the most important climate impact driver that is unambiguously associated with increasing temperatures. At the global level, warming raises sea level through thermal expansion of sea water and through melting of glaciers and ice sheets. Variations in land water storage are another important factor. At the regional scale, sea level change is influenced by large-scale ocean circulation patterns, and geologic processes and deformation from the redistribution of ice and water. Locally, vertical land motion from geologic processes, ground water withdrawal, and fossil fuel extraction are also important.

The IPCC AR6 estimates that global mean sea level increased by 7.9 (5.9–9.8) inches between 1901 and 2018, with the rate of sea level rise accelerating in recent decades. At the ocean basin scale, sea levels have risen fastest in the Western Pacific and slowest in the Eastern Pacific over the period 1993–2018 (Fox-Kemper *et al.*, 2021). The rate of global sea level rise is estimated to be 0.12 inches/year, about the height of two stacked pennies (NASA, 2020).

The observing systems for global sea level rise have advanced significantly in the satellite era, particularly with the advent of satellite altimeters in 1993. Local tide gauges have provided useful data for the past century, and even longer for a few locations. Following the end of the Little Ice Age in the mid-nineteenth century, tide gauges show that the global mean sea level began rising during the period 1820–1860, well before most anthropogenic greenhouse gas emissions.

7.2 U.S. sea level rise

Observed and predicted rates of mean global sea level rise might have little scientific relevance for specific locations, owing to local processes (NOAA, 2025). Figure 7.1 shows that in Canada and Alaska (and also northern Washington), sea level is decreasing, owing to uplift from glacial rebound. Most of the Pacific coast tide gauges show low rates of sea level rise, while largest U.S. rates are on the Gulf coast (Louisiana and Texas) and in the mid-Atlantic states (Chesapeake Bay region).

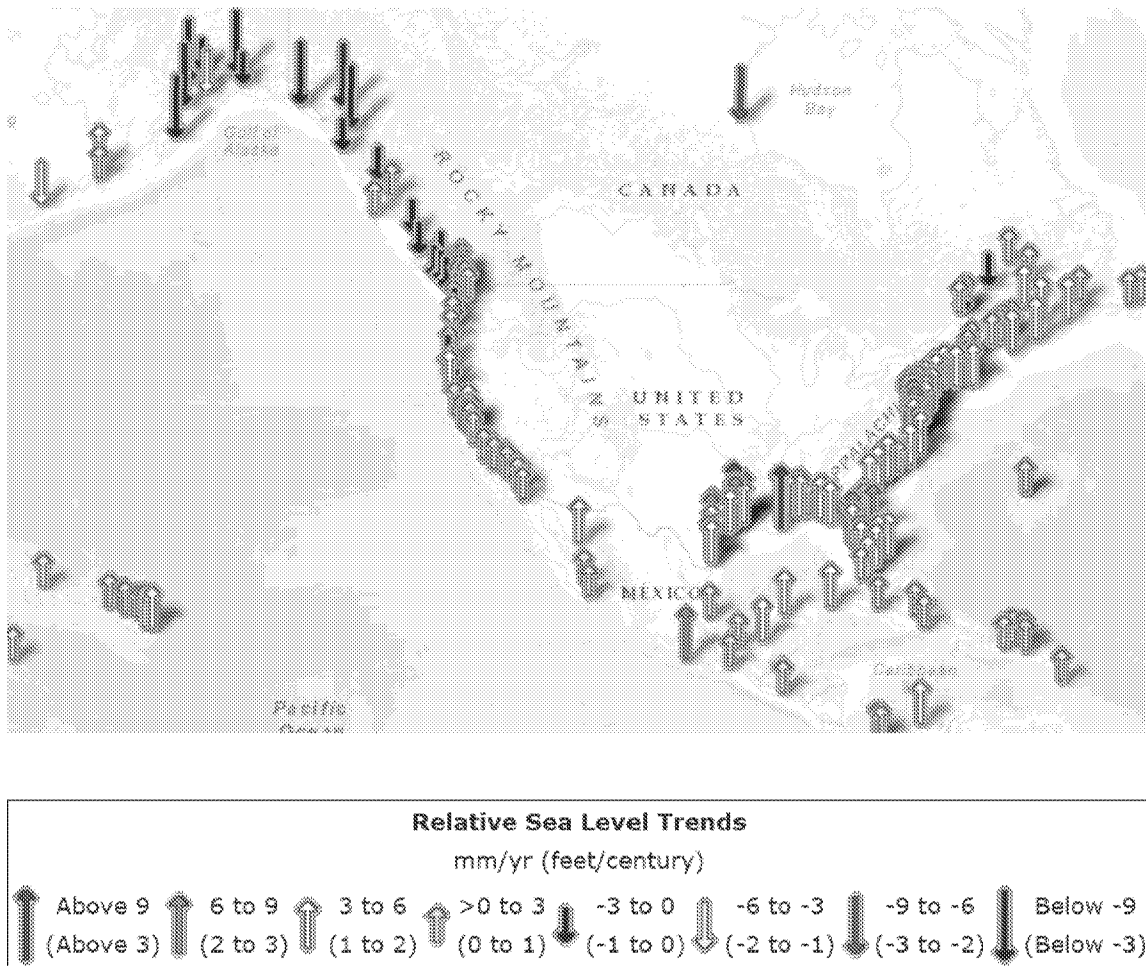


Figure 7.1. Map of rates of relative sea level rise along the U.S. coast (NOAA, <https://tidesandcurrents.noaa.gov/sltrends/>). For reference, 3 mm = 0.12 in.

Measurements of relative sea level rise from tide gauges conflate the climatologically relevant increase in the volume of seawater with local vertical land motion. The latter, which varies from place to place, is best measured using a Global Positioning System (GPS) station located near the tide gauge. It is driven by a range of processes that can be comparable to the climate signal and can locally exacerbate (subsidence/sinking) or mitigate (uplift) the risk of sea level rise (Wöppelmann and Marcos, 2016). Human activities triggering subsidence are often significant. They include soil drainage (*e.g.* for urban development) and subsurface extraction of groundwater or hydrocarbons.

Table 7.1 shows absolute sea-level rise (ASLR) for selected locations, determined from the sum of uncorrected sea-level rise as estimated from tide gauge time series (RSLR) (NOAA, 2025) and the vertical land motion (VLM) measurements (NAS, 2012; Letetrel *et al.*, 2015; Karegar *et al.*, 2016). The absolute sea level rise for each of these locations is significantly smaller than the measured relative sea level rise owing to local subsidence. More than half of the measured relative sea level rise is attributed to land sinking for these locations: San Francisco, Galveston, Grand Isle. For reference, the global average rate of absolute sea level rise is estimated to be 0.12 inches/year.

Location	RSLR	VLM	ASLR
San Francisco, CA	+0.08	-0.06	+0.02
Galveston, TX	+0.26	-0.19	+0.07
Grand Isle, LA	+0.36	-0.28	+0.08
St Petersburg, FL	+0.12	-0.02	+0.10
New York City, NY	+0.11	-0.05	+0.06

Table 7.1 Absolute sea level rise (inches/year) consisting of Relative Sea Level Rise (RSLR) plus Vertical Land Motion (VLM)

San Francisco Bay

Over the past 100 years, relative sea level in the San Francisco Bay area has risen by 7.8 inches, at an average rate of 0.08 inches/year (Figure 7.2). As shown in Table 7.1, San Francisco’s vertical land motion is -0.06 inches/year (sinking), producing a recent absolute rate of +0.02 inches/year. Portions of Treasure Island, San Francisco International Airport, and Foster City are sinking as fast as 0.4 inches/year (Shirzaei and Bürgmann, 2018). Problems in the San Francisco Bay area, including threats to the airport, are caused primarily by soil compaction in landfill zones that were formerly wetlands, not by the slow creep of global sea level rise.

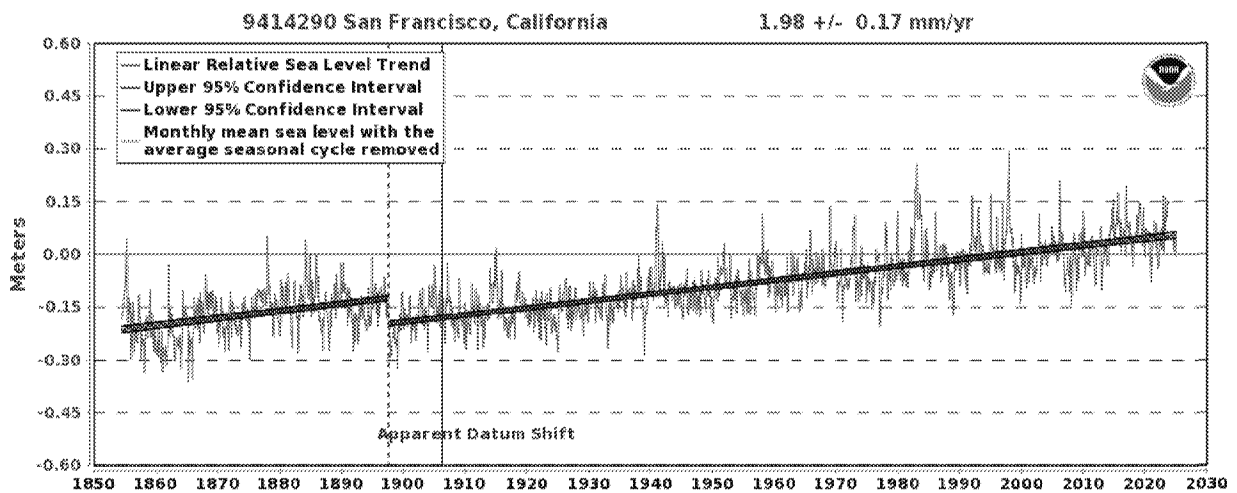


Figure 7.2. Tide gauge measurements at San Francisco, California, obtained from NOAA - https://tidesandcurrents.noaa.gov/sltrends/sltrends_station.shtml?id=9414290 (downloaded 4/22/25).

Galveston - Houston

Long-term tide measurements at Galveston Pier 21 show sea level rise of 2.18 feet in the past century, or a rate of 0.26 inches/year (Figure 7.3). Vertical land motion (subsidence) at Galveston is estimated at -0.19 inches/year, yielding an absolute rate of rise of +0.07 inches/year (Table 7.1). The U.S. Geologic Survey found that most of the land-surface subsidence in the Houston-Galveston region is a direct result of groundwater withdrawals (Kasmarek and Ramage 2017), which caused compaction of the aquifer sediments, mostly in the fine-grained silt and clay layers. By 1979, as much as 10 feet of subsidence had occurred in Houston.

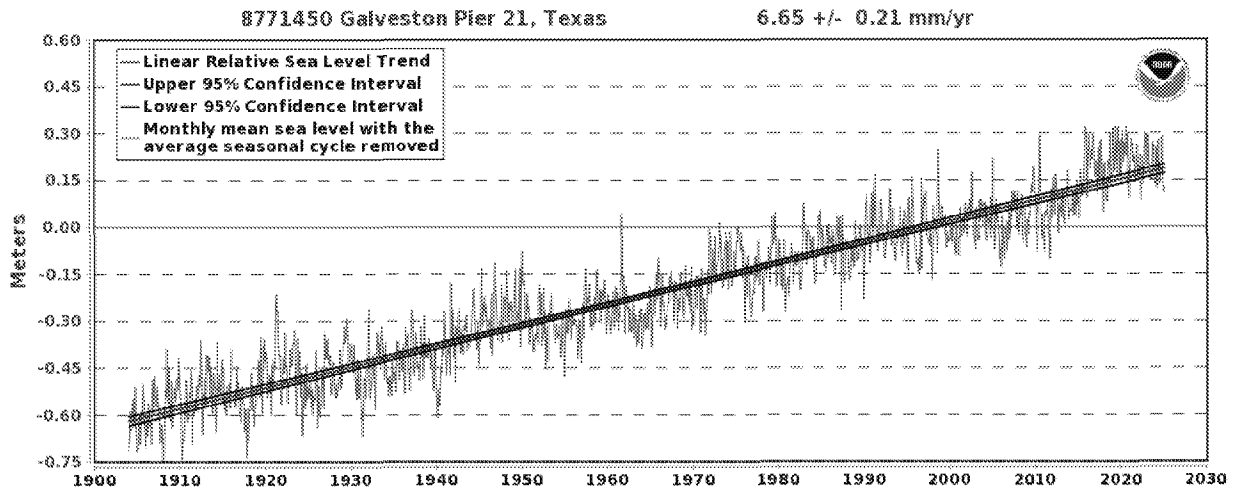


Figure 7.3. Tide gauge measurements Galveston Pier, TX, obtained from NOAA - https://tidesandcurrents.noaa.gov/sltrends/sltrends_station.shtml?id=8771450 (downloaded 4/22/2025).

New Orleans and the Mississippi delta

Long-term tide gauge measurements at Grand Isle, Louisiana, show that sea level has risen by slightly more than 3 feet over the last 100 years at an average rate of 0.36 inches/year (Figure 7.4). Vertical land motion (subsidence) is estimated -0.28 inches/year. Table 7.1 gives the absolute sea level rise as +0.08 inches/year.

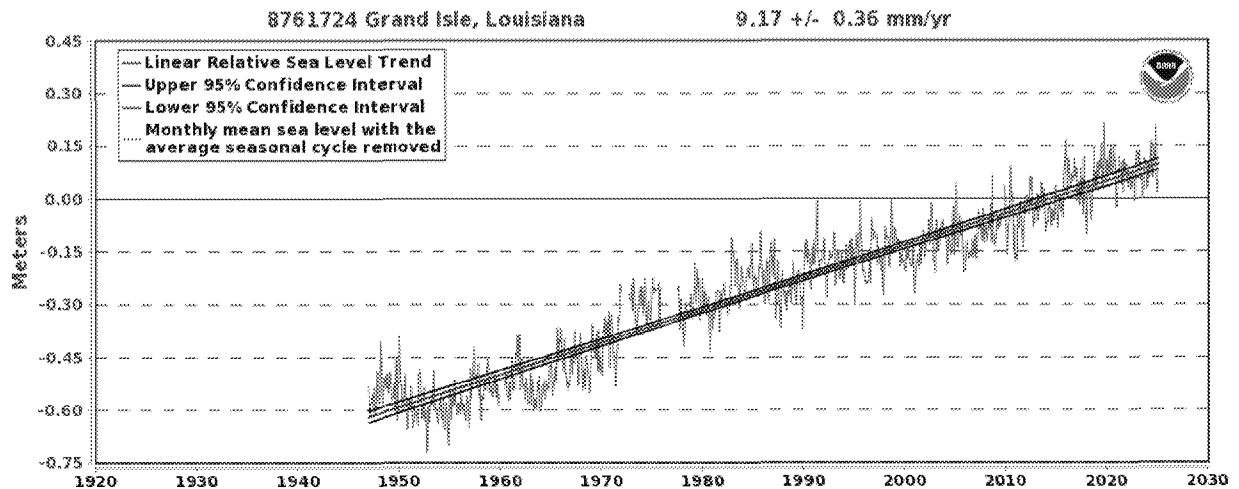


Figure 7.4. Tide gauge measurements at Grand Isle, LA, obtained from NOAA - https://tidesandcurrents.noaa.gov/sltrends/sltrends_station.shtml?id=8761724 (downloaded 4/22/25).